



Cellular Automata Models of Traffic on Ant Trails

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Introduction

Organized traffic not only in human systems !!

- different forms of traffic in other biological systems
- similarities and differences to human traffic?
- optimized by evolution

biology: "collective intelligence" interactions between units of low intelligence can create behaviour that appears to be "intelligent"

- physics: general aspects (non-equilibrium!)
 - collective phenomena

Asymmetric Simple Exclusion Process

Cellular Automata

Cellular automata (CA) are discrete in

- space
- time
- state variable (e.g. occupancy, velocity)



Advantage:

- dynamics in form of intuitive "rules"
- efficient implementation for large-scale computer simulations

often: stochastic dynamics

Asymmetric Simple Exclusion Process

Caricature of traffic:

Asymmetric Simple Exclusion Process (ASEP):

- 1. directed motion
- 2. exclusion (1 particle per site)
- 3. stochastic dynamics

For applications: different modifications necessary



ASEP

ASEP = "Ising" model of nonequilibrium physics

- simple
- exactly solvable
- many applications

Applications:

- Protein synthesis
- Surface growth
- Traffic
- Boundary induced phase transitions

Periodic boundary conditions

fundamental diagram



no or short-range correlations

Influence of Boundary Conditions

open boundaries: density not conserved!



exactly solvable for all parameter values!



Dynamics on Ant Trails

Ant Trails

ants build "road" networks: trail system



Chemotaxis

Ants can communicate on a chemical basis:

chemotaxis

Ants create a chemical trace of pheromones

trace can be "smelled" by other ants follow trace to food source etc.



Chemotaxis



chemical trace: pheromones

Ant Trail Model

Basic ant trail model: ASEP + pheromone dynamics

- ants create pheromones
- hopping probability depends on density of pheromones
- distinguish only presence/absence of pheromones
- 'free' pheromones evaporate

Ant Trail Model

1. motion of ants

Dynamics:

2. pheromone update (creation + evaporation)



parameters: q < Q, f

Fundamental diagram of ant trails

velocity vs. density



non-monotonicity at small evaporation rates!!

Experiments: Burd et al. (2002, 2005) John et al. (2006)

different from highway traffic: no egoism

Spatio-Temporal Organization

dynamical disorder: formation of "loose clusters"



early times

steady state

coarsening dynamics: cluster velocity ~ gap to preceding cluster

Traffic on Ant Trails



Formation of clusters

Mapping on Zero-Range Process



Mapping on Zero-Range Process

Approximation: hopping rates depend only on headway

$$p(T) = q + (Q - q)(1 - f)^{T}$$

$$T = \frac{d}{v}$$

(T=time since pheromone has been created)

(d=distance headway, v=velocity)

 \rightarrow in equivalent ZRP: hopping rate p(n) (n=d)

requires self-consistent determination of velocity v

Mapping on Zero-Range Process

Zero-range process exactly solvable ! product measure (mean-field exact)

good agreement for fundamental diagrams!

Phase transition ??

only in limit $f \to 0$ (does not commute with limit $L \to \infty$)

Counterflow

model: 2 separate lanes with common pheromone trail





hindrance effect through interactions (e.g. for communication)

K < Q

Counterflow



localized clusters are formed !

Ant Trails – Empirical Results

Field Studies: Setup

Data Collection

- stable trail with single-lane traffic (uni- or bidirectional)
- suitable climate (excluding heavy rainfall or very high temperatures)

Try to find simple situations excluding additional effects

- observe the bulk of the trail (neglect boundary effects)
- avoid intersections or branching (conservation of particles)
- preferably no obstacles like stones or leaves (translational invariant trail)





(A. John, Dissertation 2007)

Field Studies: Basic Quantities

- -basic quantities are extracted from cumulative counting (t(n),n)
- -based on the assumption: n-th ant in is also the n-th ant out

1 2 3 4 5 B

Travel Time and single-ant velocity:

$$dT(n) = t_{-}(n) - t_{+}(n)$$
$$v(n) = \frac{L}{dT(n)} = \frac{L}{t_{-}(n) - t_{+}(n)}$$

Instantaneous particle number:

$$N(t) = n_{+}(t) - n_{-}(t)$$



Field Studies: Empirical Results

Fundamental Diagrams

- no overtaking
- average velocity shows only weak density dependence
- slight decrease of average velocity leading to non-linear increase of flow

Main Observation:

- mutual blocking seems to be suppressed (no congested state)





Field Studies: Empirical Results

Distance-Headway Distribution

- characterizes the spatial distribution of ants on the trail
- two regimes can be distinguished

1. Intra-Platoon (log-normal)

$$P(d) = \frac{1}{d\sigma_L \sqrt{2\pi}} \exp\left(-\frac{(\mu_L - \log(d))^2}{2\sigma_L^2}\right)$$

2. Inter-Platoon (random-headways)

$$P(d) = \frac{1}{\lambda} \exp\left(-\frac{d}{\lambda}\right)$$



Field Studies: Empirical Results

Velocity Distribution

- Gaussian distribution with slight asymmetry

$$P(v) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{(v(n) - V)^2}{2\sigma^2}\right)$$



Density Dependent Distributions:



Summary

- Ant form trail system through chemotaxis
- Chemotaxis also influences dynamics on existing trails
- model based on extended ASEP
- induces tendency of clustering
- fundamental diagram shows non-monotonic behaviour
- empirical studies:
 - clustering
 - velocity density-independent
 - narrow velocity distribution
- Pedestrian dynamics (virtual chemotaxis)

Collaborators

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