

The Evolution of Animal Grouping and Collective Motion

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COLLECTIVE

ANIMAL

BEHAVIOR



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Courtesy: Iain Couzin/BBC

Broad relevance of collective behavior

- Observed in wide range of organisms and ecological contexts
 - From bacterial (or cellular) swarms to wildebeests
- Nonequilibrium statistical mechanics of self-driven particles
- Robotics: Coordination of mobile autonomous agents
- Natural Algorithms
- Traffic Organization, Human Crowds, etc.

Question: Individuals to collectives

1. How do individual level interactions scale to collective patterns

Individual level interactions



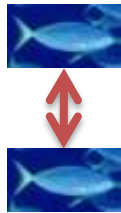
Self-ordered collective motion



Why such interactions occur?

1. How individual level interactions scale to collective patterns

Individual level interactions



Self-ordered collective motion



- **Local interactions can change**

- Over relatively **long evolutionary timescales** (due to reproduction, mutation and natural selection)

2. Why do such interactions among individuals occur/exist in the natural world?

Individual versus collective benefits

- One may argue that local social interactions lead to emergent group benefits (or mutual benefits to individual members).
 - *e.g.*, Improved navigational and/or foraging ability
- But natural selection does not optimize group properties.
 - It favors individuals with higher relative-fitness, typically leading to conflict among individuals.
- What happens in a migratory context?

Collective motion in migratory species

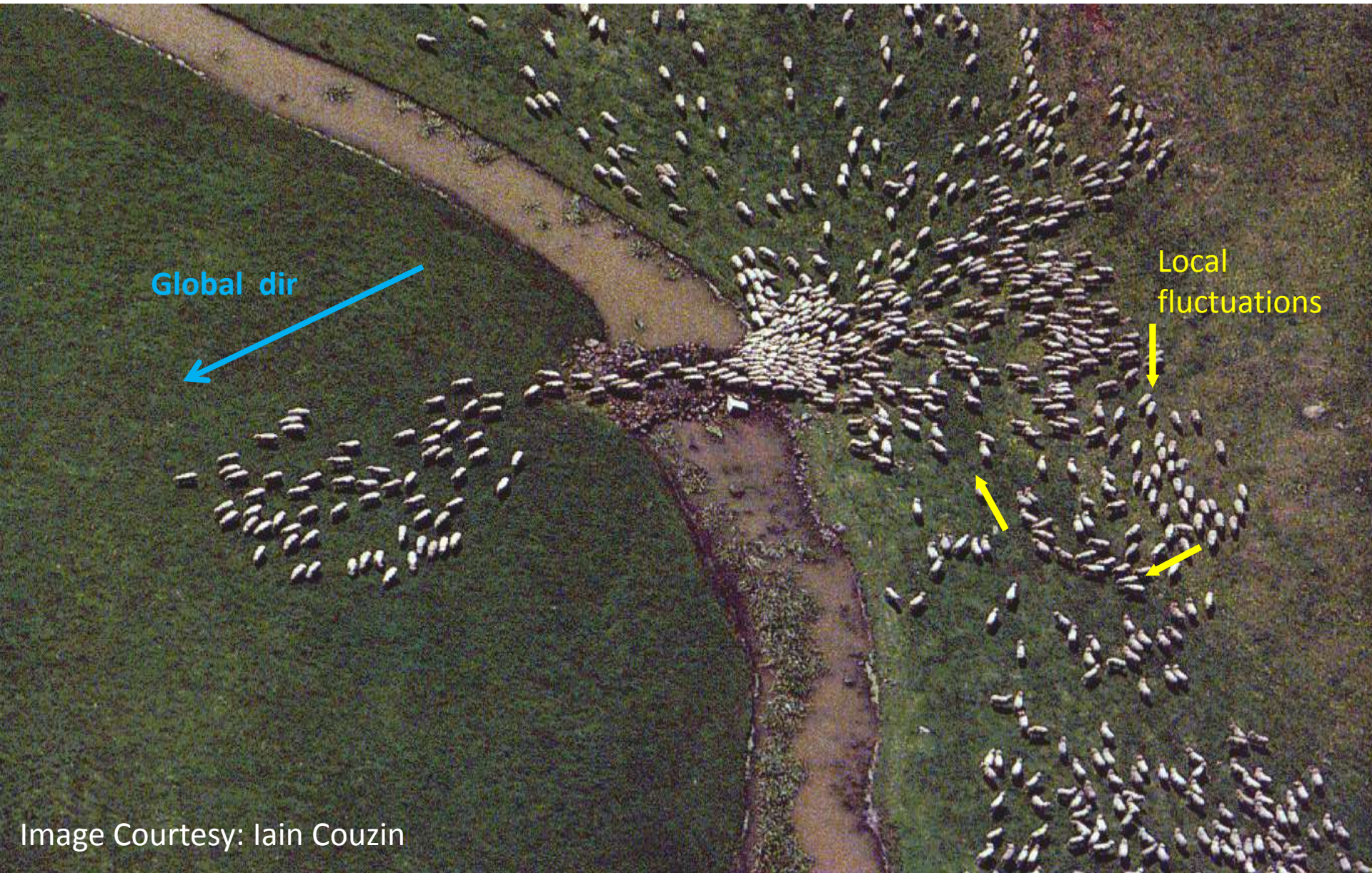
- Ubiquitous phenomena: Cells to Wildebeests.
 - Involves climbing gradients (magnetic, resource, etc).
- Massive number of individuals



Image credits:

(L) Tamas Vicsek, (C) http://forces.si.edu/images/el_nino/locustLarge.jpg, (R) http://www.martinwgrosnick.com/images/Wildebeest_48284F.jpg

Collective migration: hints on mechanism



Global dir

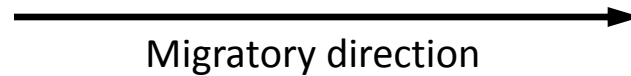
Local
fluctuations

A self-propelled particle model for collective migration

- We will consider following two individual traits
 - Ability to sense environmental gradient/cues
 - Geomagnetic field, resource, thermal, chemical and/or electromagnetic fields.
 - Ability to socially interact with neighbors
 - Respond to neighbors motion, *e.g.*, through visual or chemical cues.
- Given a need to migrate in a specific direction:
 - How do individuals optimize above two traits.
- Individuals (self-propelled particles) move
 - in a continuous two-dimensional space
 - with constant speed
 - update their direction of motion in discrete time steps.

Solitary migration

- Environment: a global migratory gradient along x-axis.

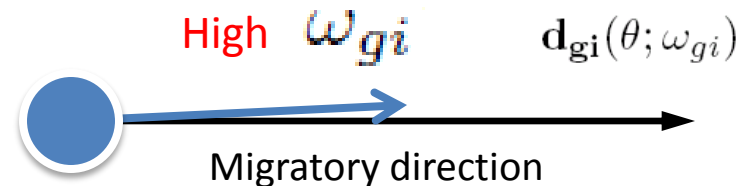
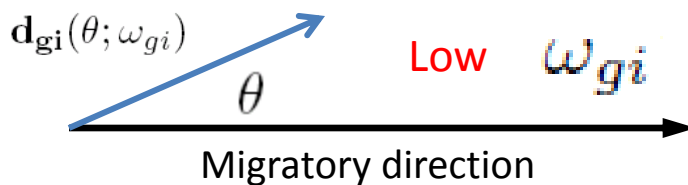


- Trait 1: Gradient detection trait (ability): ω_{gi}

- Determines how accurately an individuals find migratory direction.

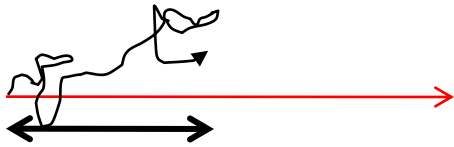
$$\dot{\theta}_i = -\omega_{gi}\theta_i(t) + \sigma_{\eta}\eta_{gi}(t)$$

- For simplicity, assume temporally uncorrelated noise.



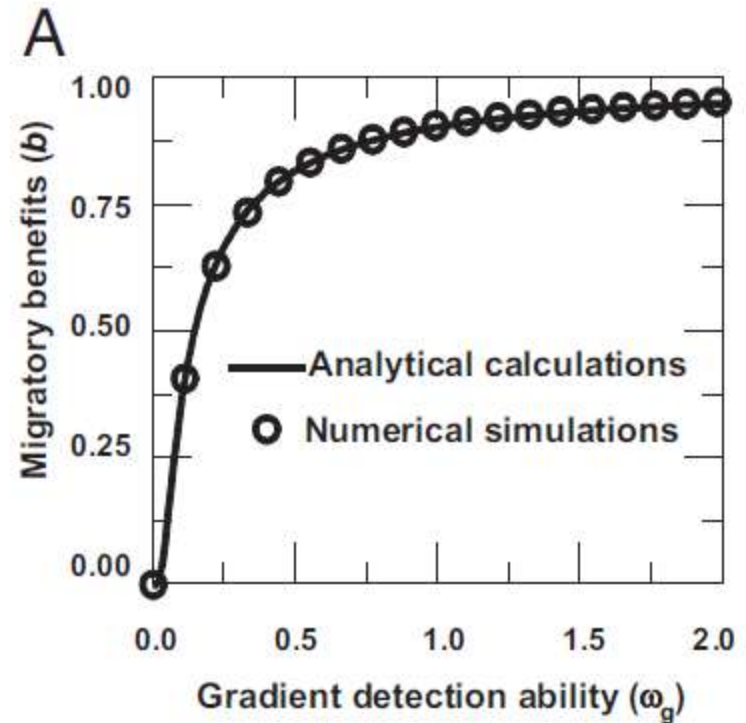
Migratory benefits

- b_i = distance moved in the direction of gradient, per unit time



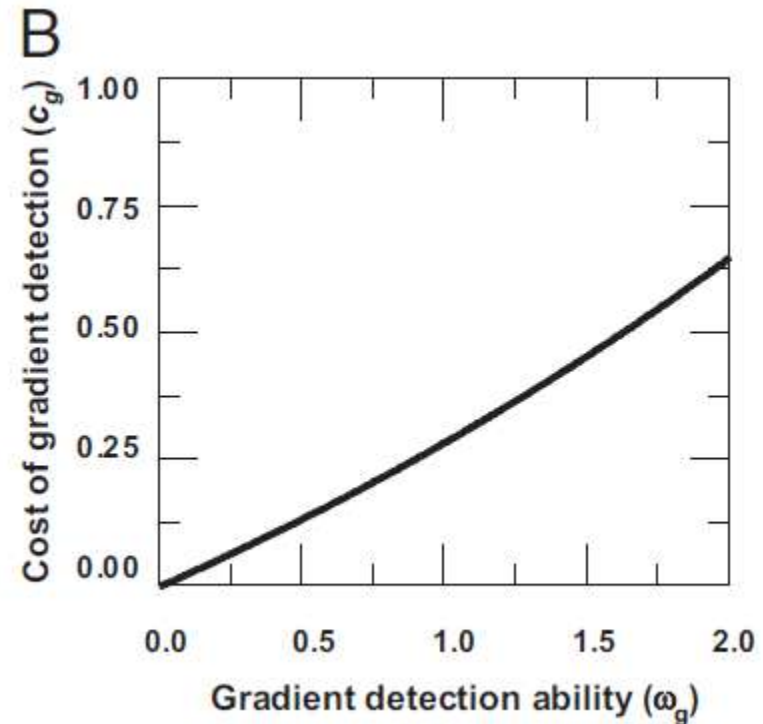
$$\dot{\theta}_i = -\omega_{gi}\theta_i(t) + \sigma_g\eta_{gi}(t)$$

$$\begin{aligned}\langle b_i \rangle &= \langle \cos(\theta_i(t)) \rangle \\ &= \exp\left[-\frac{\sigma_g^2}{4\omega_{gi}}\right]\end{aligned}$$



Costs of gradient detection ability

$$c_{gi} = p_g(e^{\omega_{gi}/\omega_{gc}} - 1.0)$$

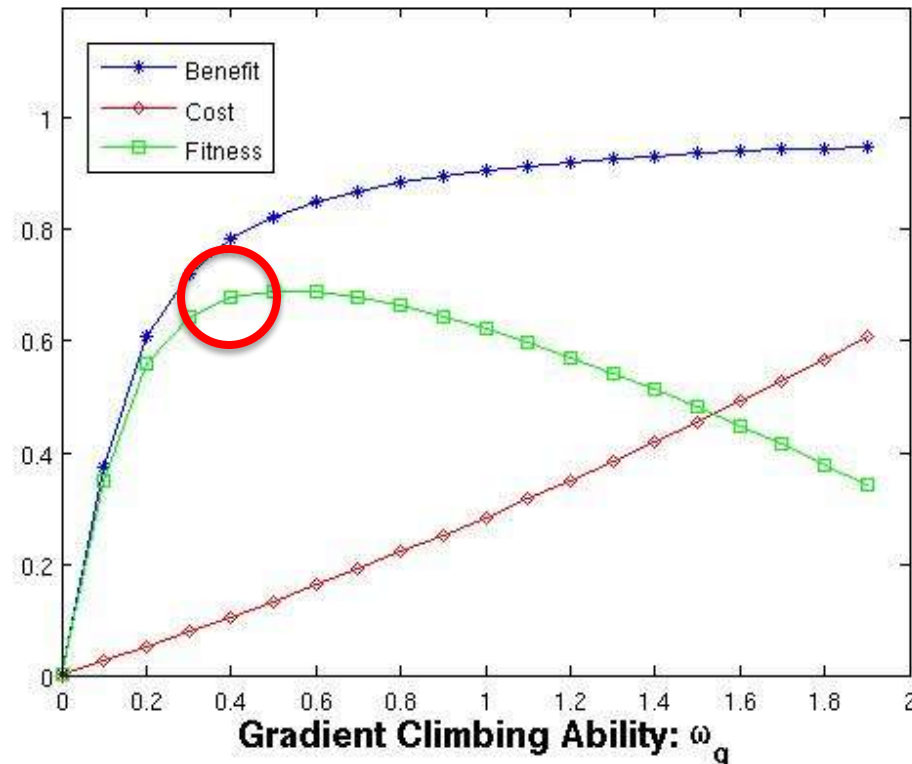


Note:

- p_g has same units as migratory benefits.
- Different magnitude of costs is obtained by varying p_g

Individual migratory fitness

- Fitness (reproductive success) = Benefit – Cost.



- Solitary individual optimum.
- What if we allow flocking interactions between individuals?

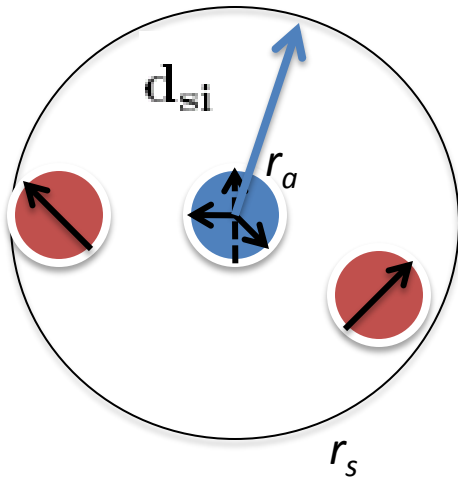
Collective motion

- Trait 2: Flocking/ Social interactions: ω_s

Denote position by $\mathbf{c}_i(t)$ and velocity by $\mathbf{v}_i(t)$

Repulsion: Avoid others within a body length r_a

Sociality: a tendency to attract towards, and align direction of travel, with neighbors within radius r_s (outer radius)



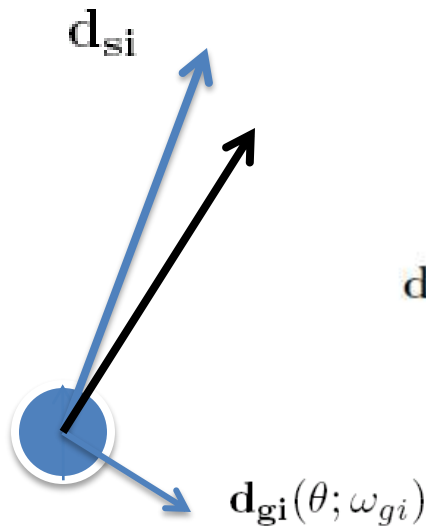
$$\mathbf{d}_{si}(t + \Delta t) = \underbrace{\sum_{j \neq i} \frac{\mathbf{c}_j(t) - \mathbf{c}_i(t)}{|\mathbf{c}_j(t) - \mathbf{c}_i(t)|}}_{\text{Attraction}} + \underbrace{\sum_{j=1} \frac{\mathbf{v}_j(t)}{|\mathbf{v}_j(t)|}}_{\text{Alignment}}$$

V Guttal and I D Couzin, 2010, *PNAS*, Vol: 107: 16172-16177

Based on Couzin *et al*, Nature, 2005

Balancing two traits/tendencies

- Gradient detection trait (ability): ω_g
- Sociality trait: ω_s



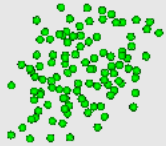
$$d'_i(t + \Delta t) = \frac{\omega_{si} \hat{d}_{si}(t + \Delta t) + \omega_{gi} \hat{d}_{gi}(t + \Delta t) + \sigma_r \hat{d}_{ri}(t + \Delta t)}{|\omega_{si} \hat{d}_{si}(t + \Delta t) + \omega_{gi} \hat{d}_{gi}(t + \Delta t) + \sigma_r \hat{d}_{ri}(t + \Delta t)|}$$

Error in perception and/or motion

V Guttal and I D Couzin, 2010, *PNAS*, Vol: 107: 16172-16177

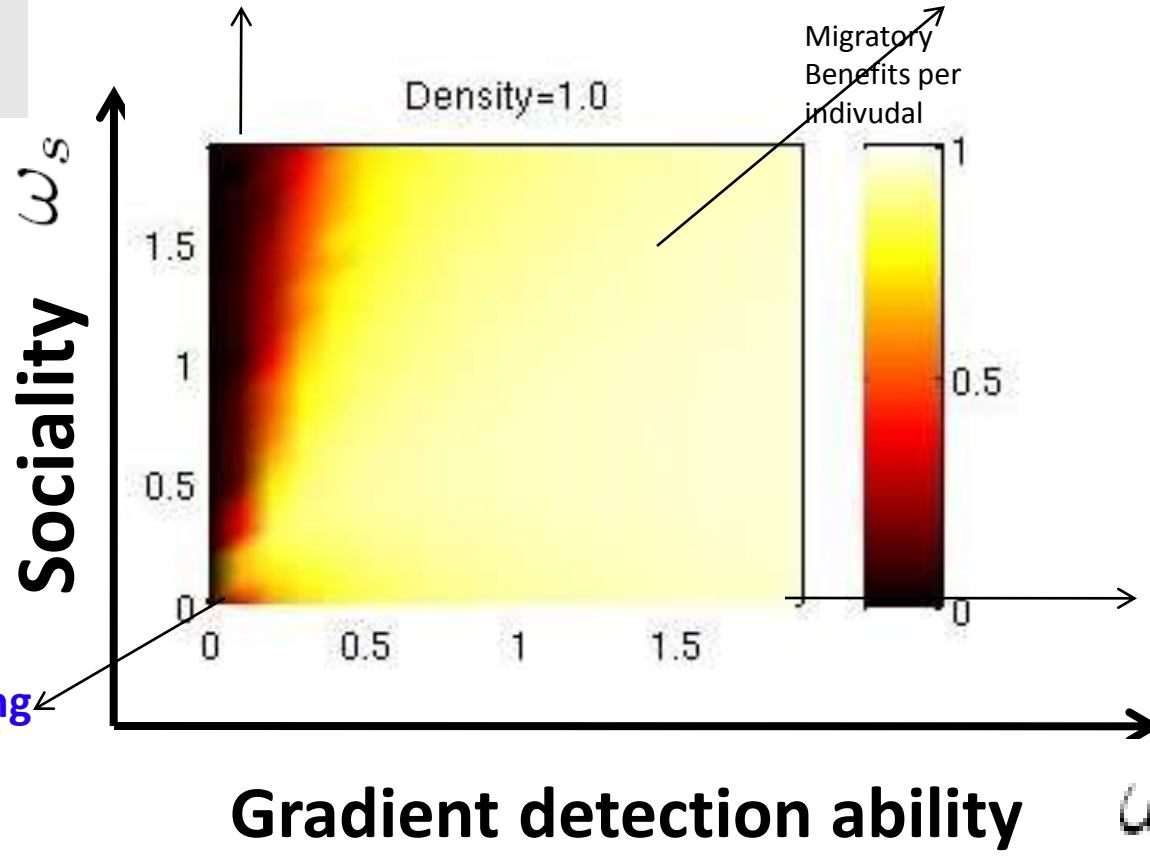
Based on Couzin *et al*, Nature, 2005

Plausible population level dynamics



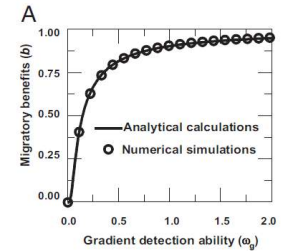
Brownian swarms

Collective Migration



Random walking individuals

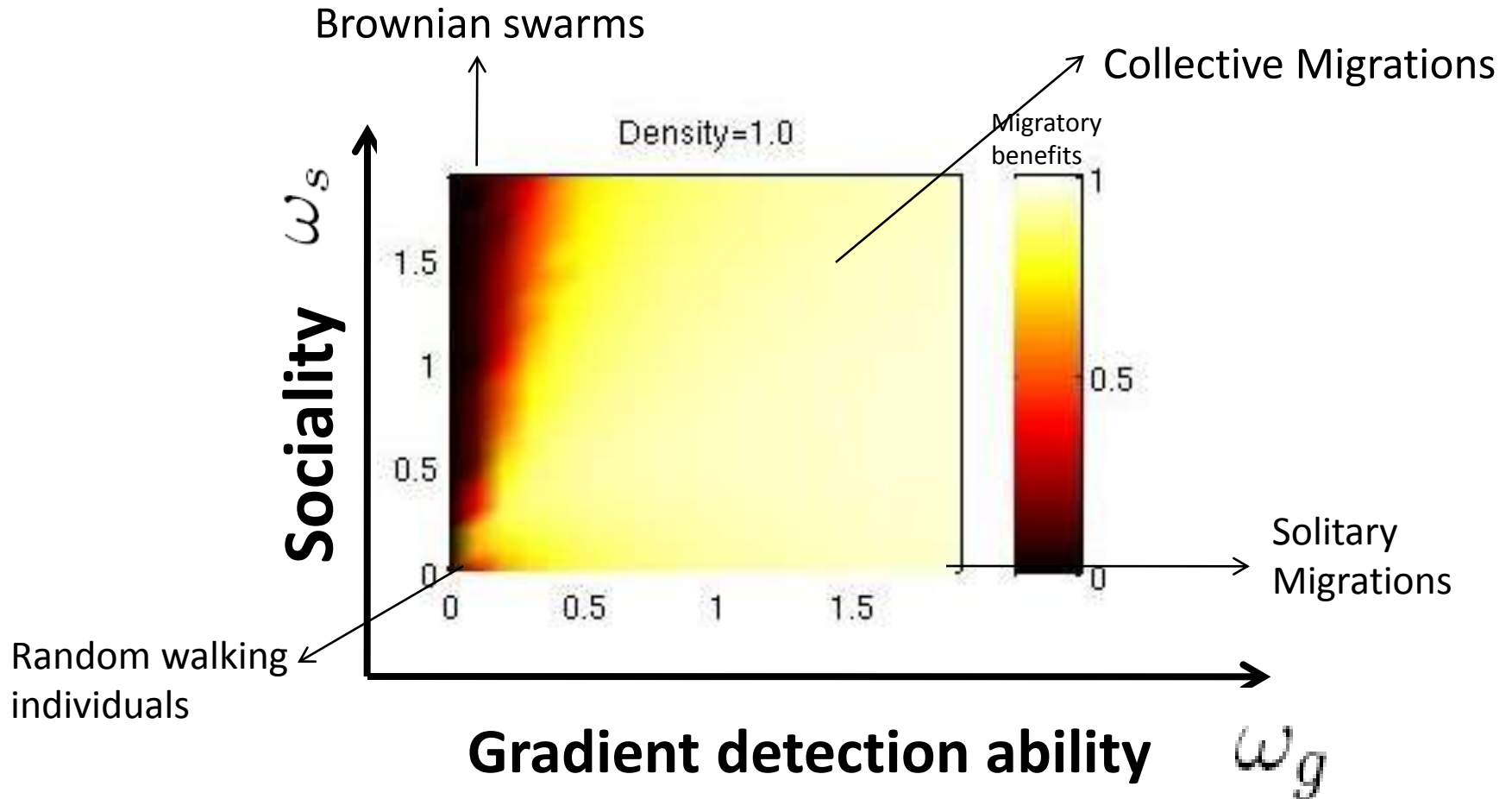
Solitary Migrations



Individual traits are determined by natural selection

- We do not predetermine what the individual traits in the population are.
- It is going to be determined by natural selection.
 - Individuals may possess only gradient detection ability.
 - Or only social interactions and thus, follow nearby individuals.
 - Or a combination of both.
 - Evolved populations may also be heterogeneous.

Plausible population level dynamics



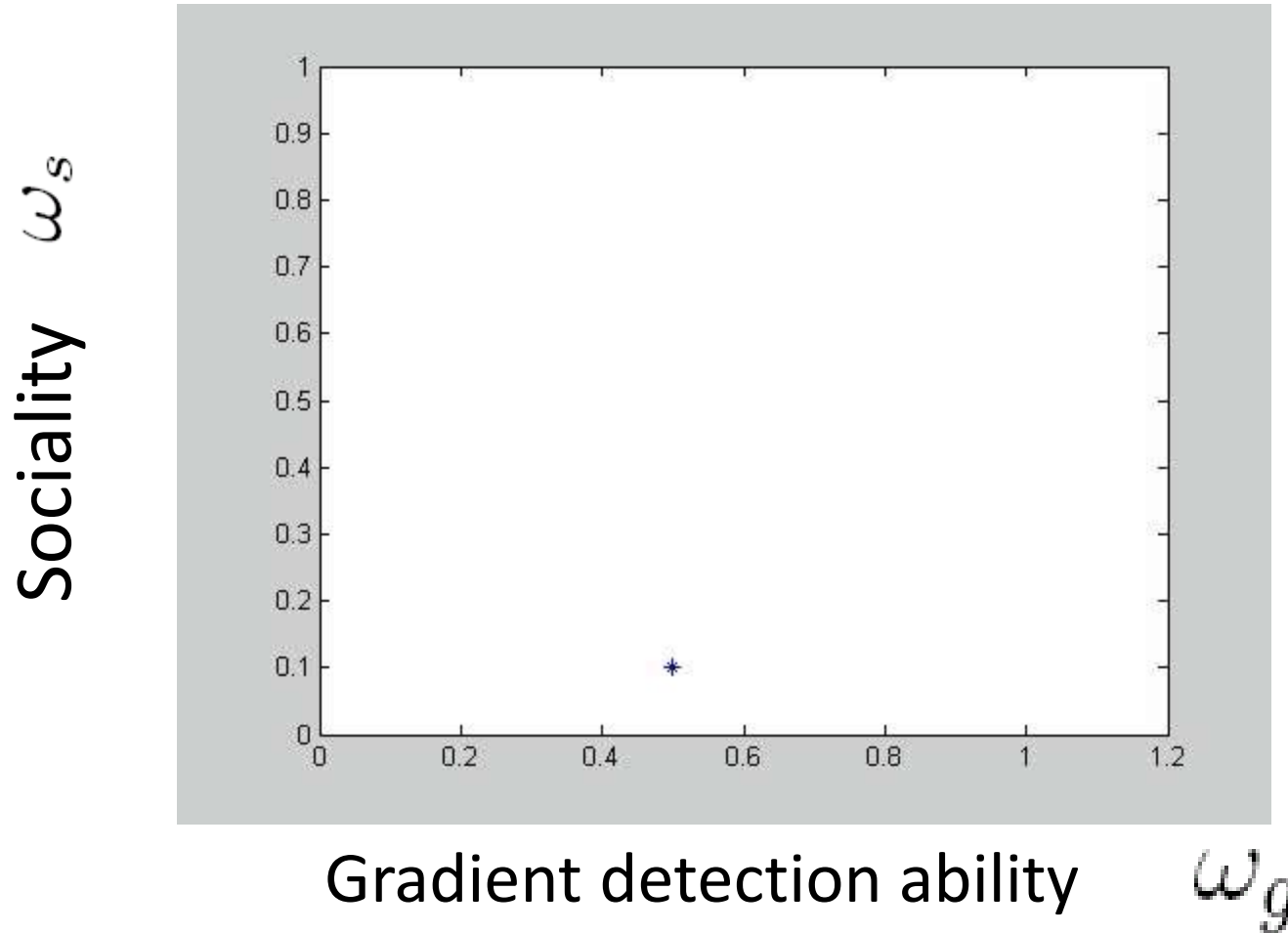
- Where would evolved populations be in this parameter space?

Selection algorithm

- Start with a large homogeneous population (large sizes: 16,000 to 60,000 individuals)
- They move according to the equations of motion over one generation (averaged over several realizations) .
- Calculate Fitness:
 - Find benefit for each individual
 - Find cost of gradient detection depending on its ω_g
 - Calculate Fitness = Benefit – Cost.
- Reproduce proportional to fitness (Roulette Wheel Selection)
 - Asexual reproduction
 - The gradient detection ability, ω_g and the sociality trait, ω_s of the parent are passed on to offspring with a small mutation.
- Repeat the process until an equilibrium of trait/phenotype distributions is reached.

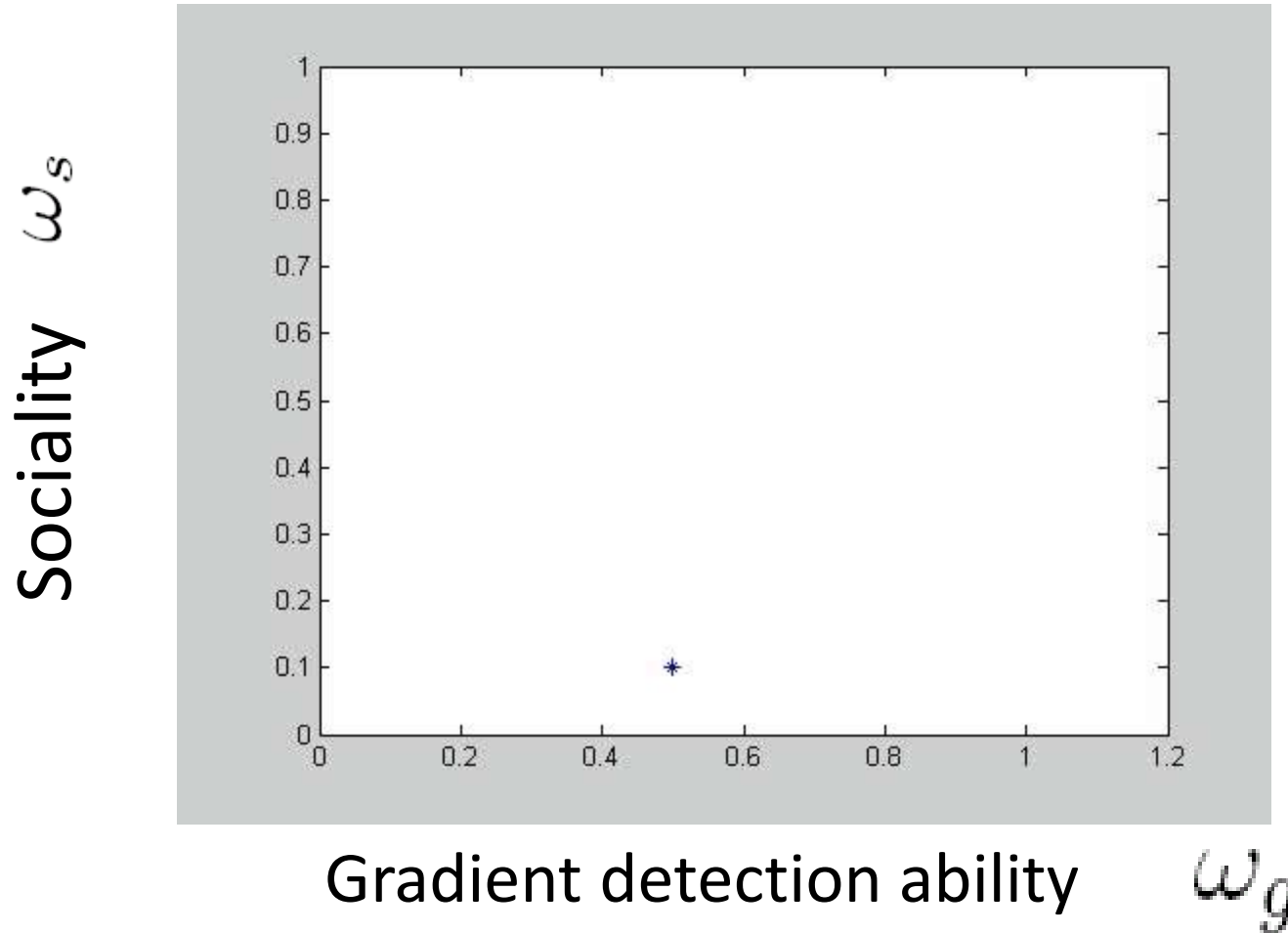
Evolved strategies:

- No cost of gradient climbing:



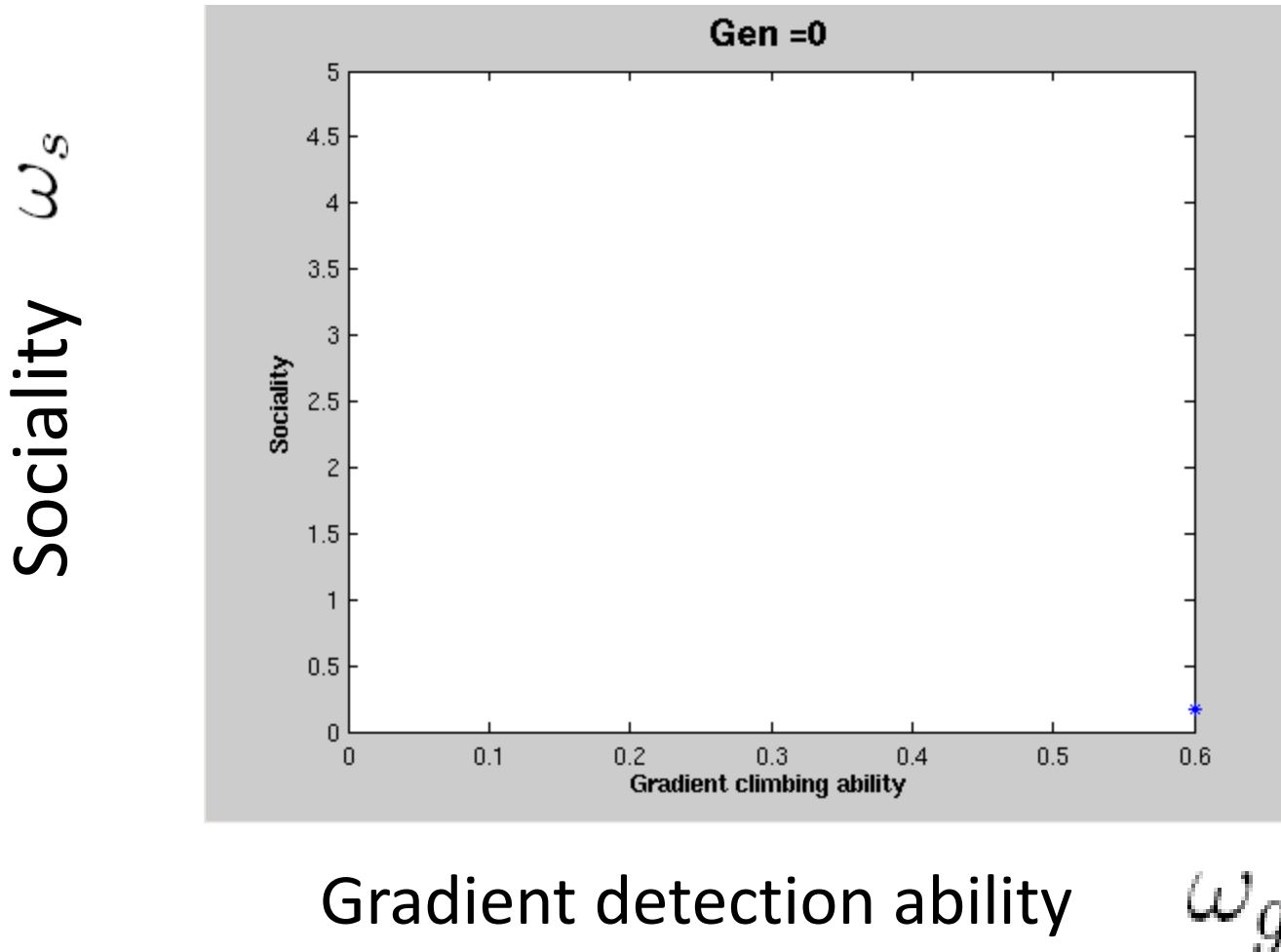
Evolved strategies:

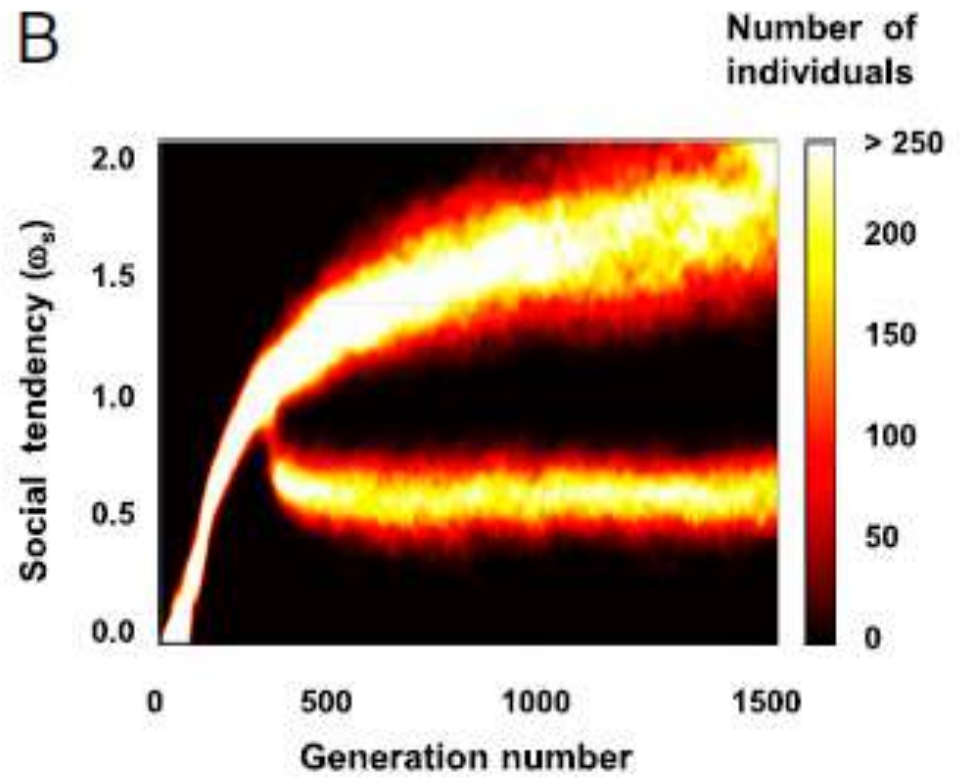
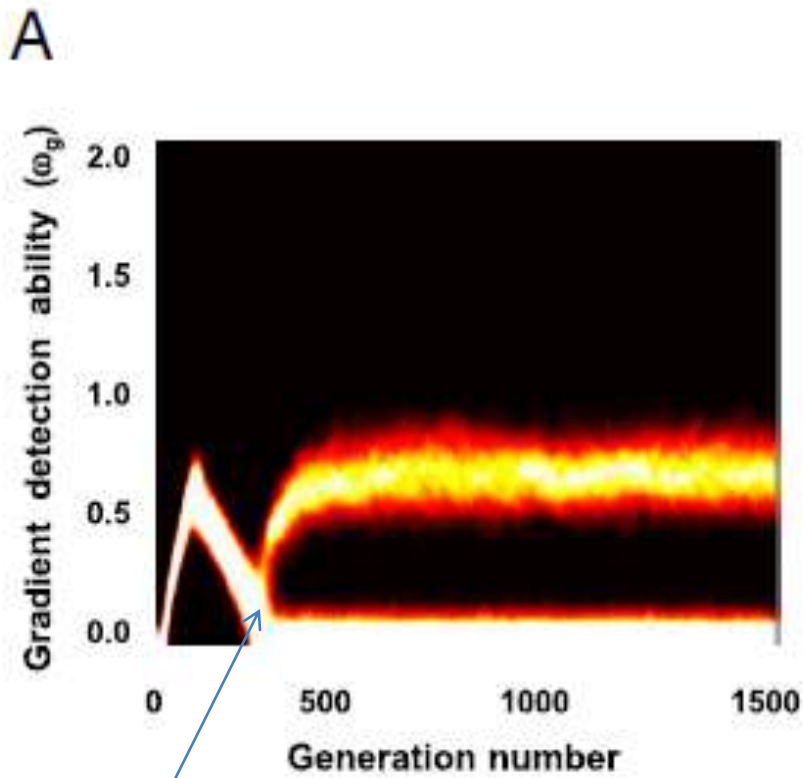
- Very high cost of gradient climbing:



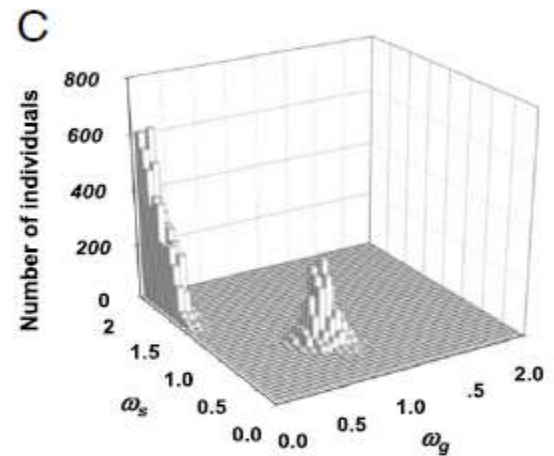
Evolved strategies:

- Intermediate cost of gradient climbing:



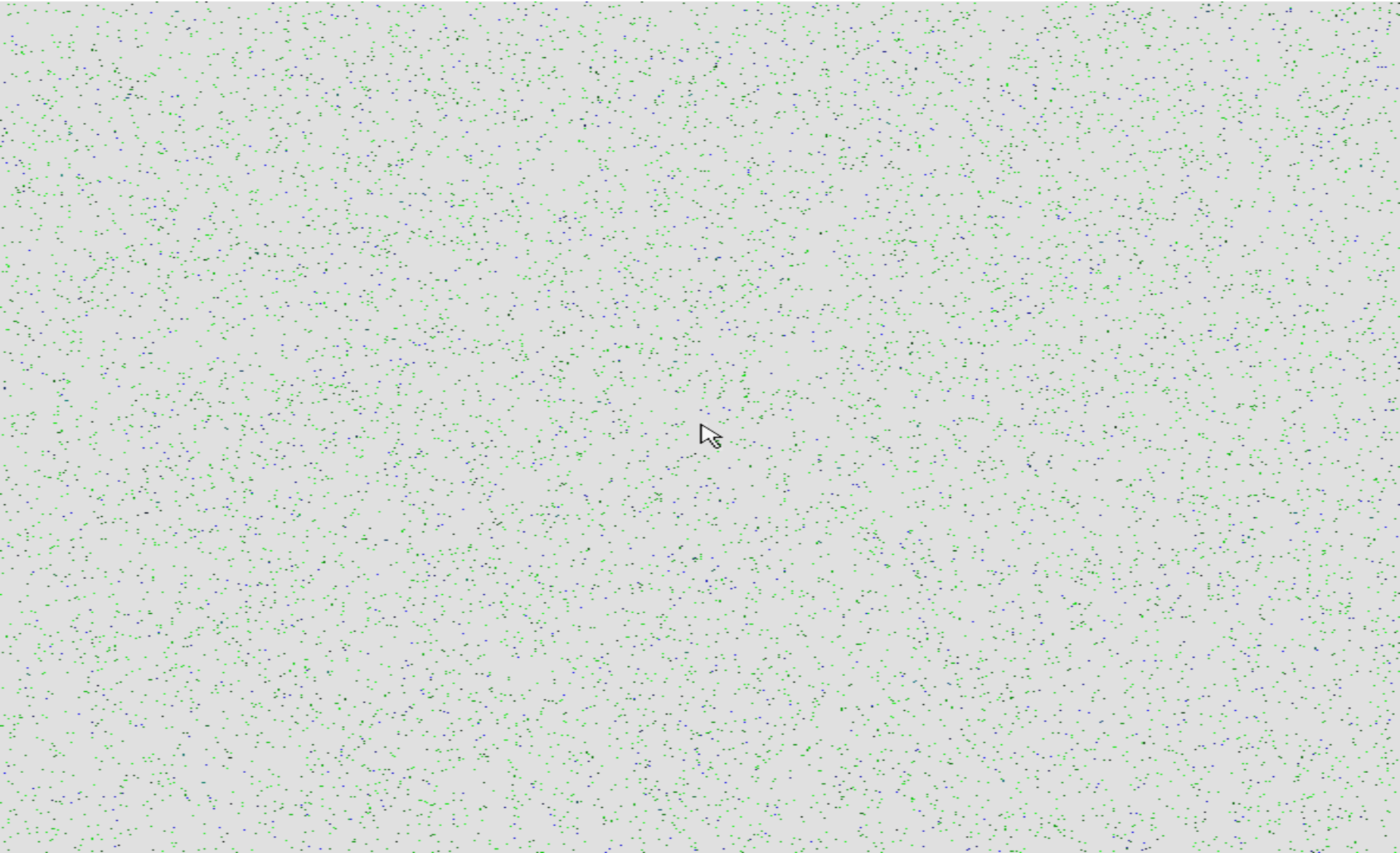


Evolutionary branching process

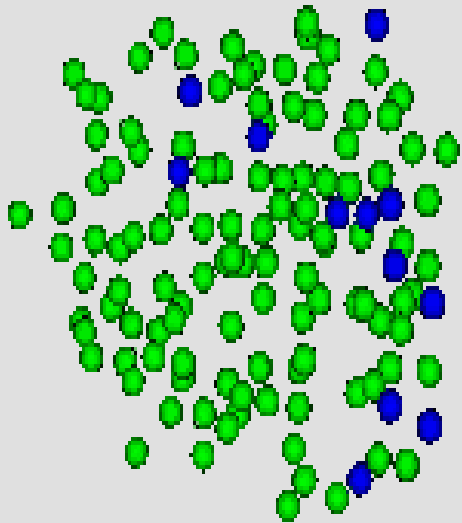


Spatiotemporal dynamics of the evolved

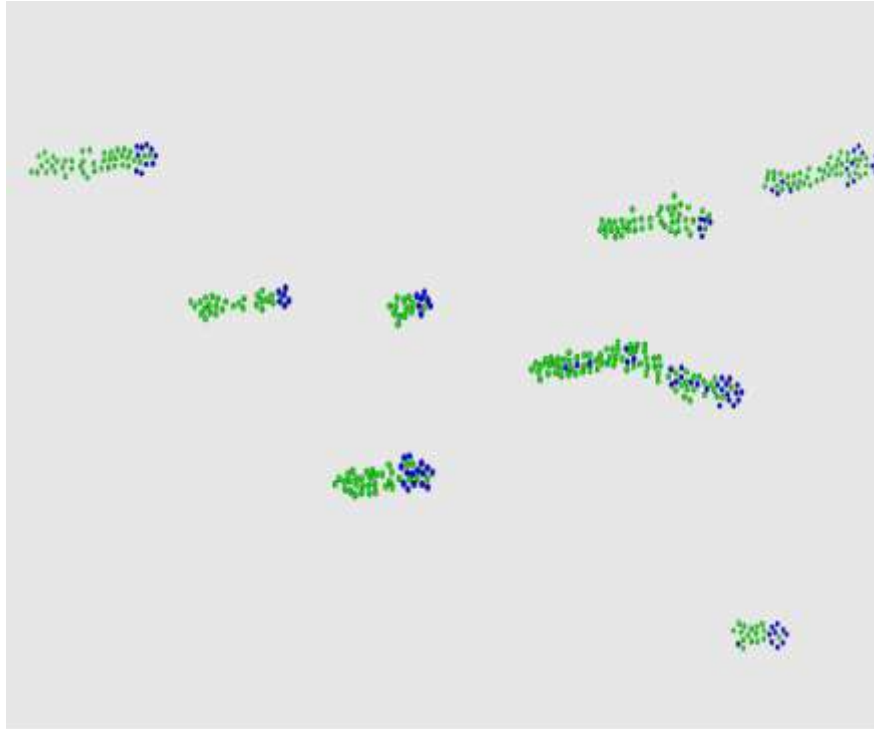
population: 16,000 individuals using GPU (CUDA) on this laptop (realtime)



Self-sorting and collective migration



Group structure in the evolved population

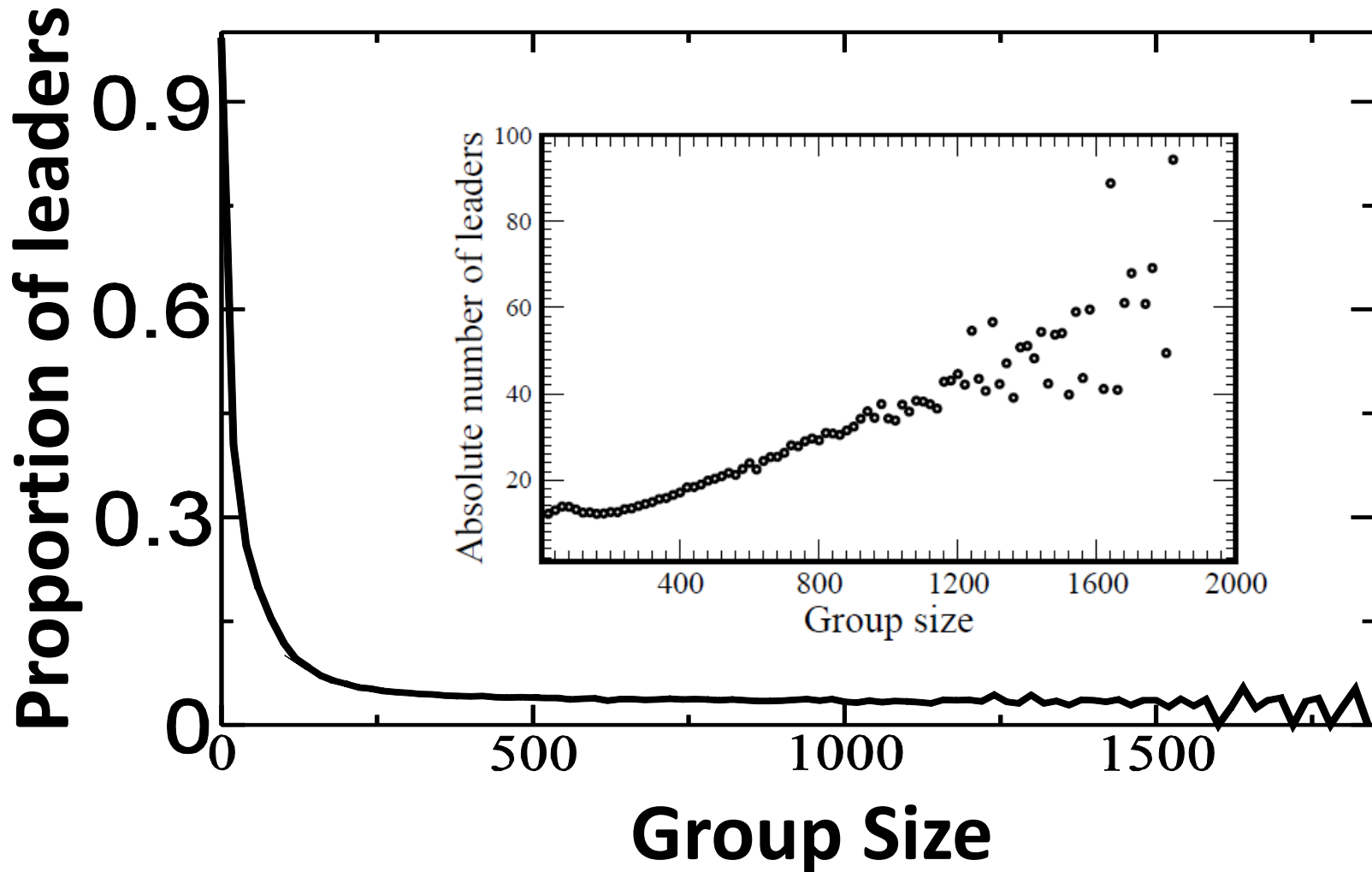


GREEN: Social individuals (high sociality but weak or no gradient detection ability)

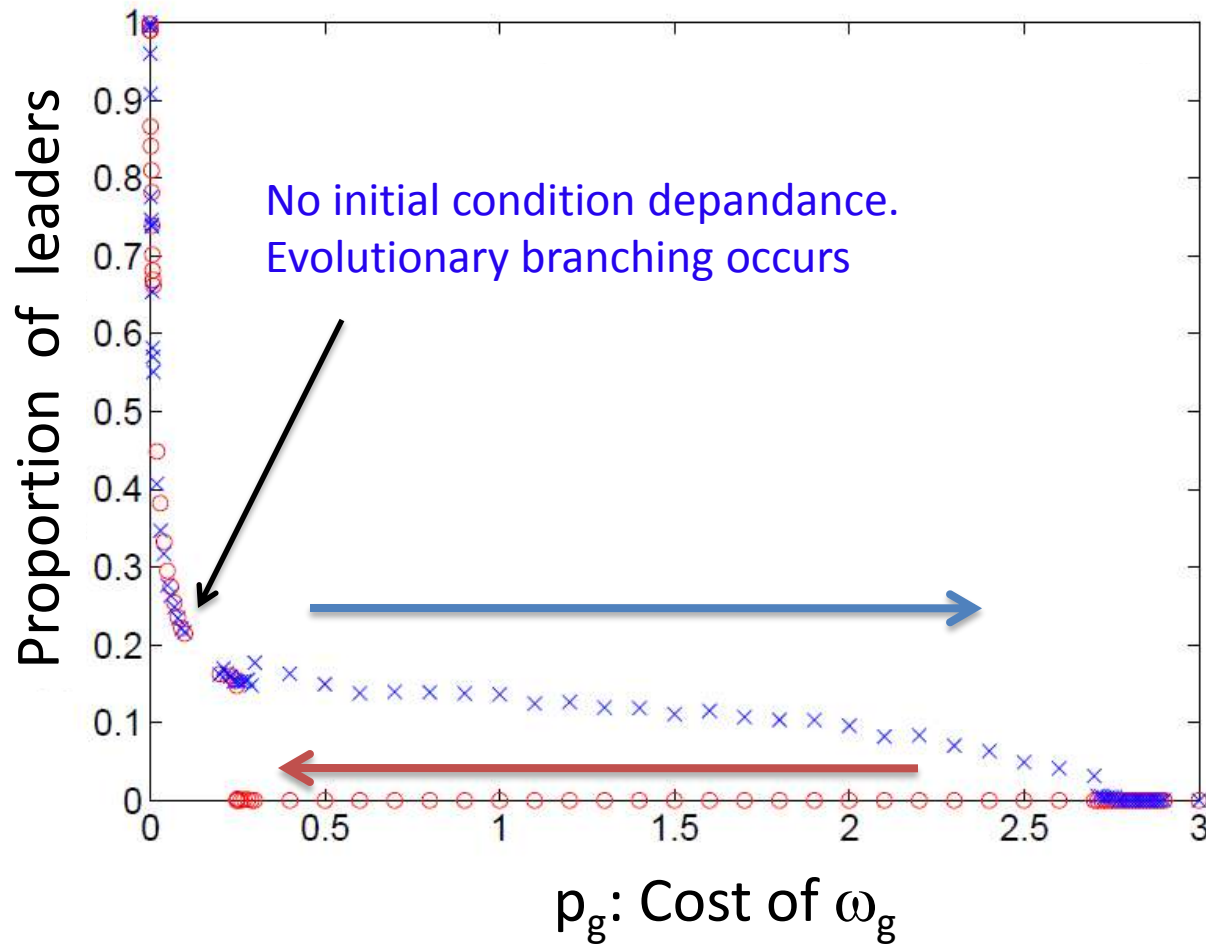
BLUE: Leaders (high gradient climbing ability and low sociality)

Self-organize into groups with mixture populations and they collectively migrate

Group structure in the evolved population

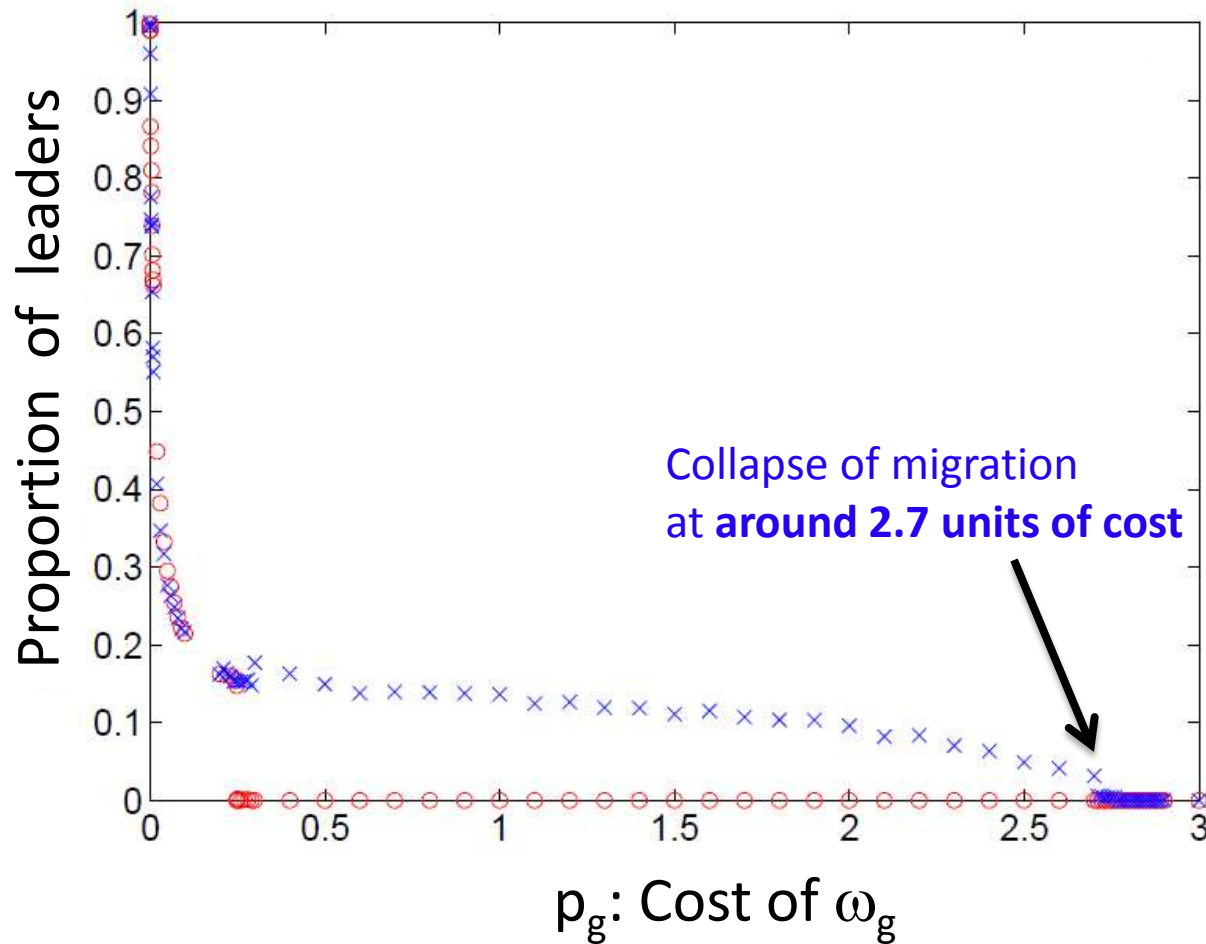


Strategies as a function of *cost of ω_g*



$$c_{gi} = p_g (e^{\omega_{gi}/\omega_{gc}} - 1.0)$$

Solitary versus social strategies

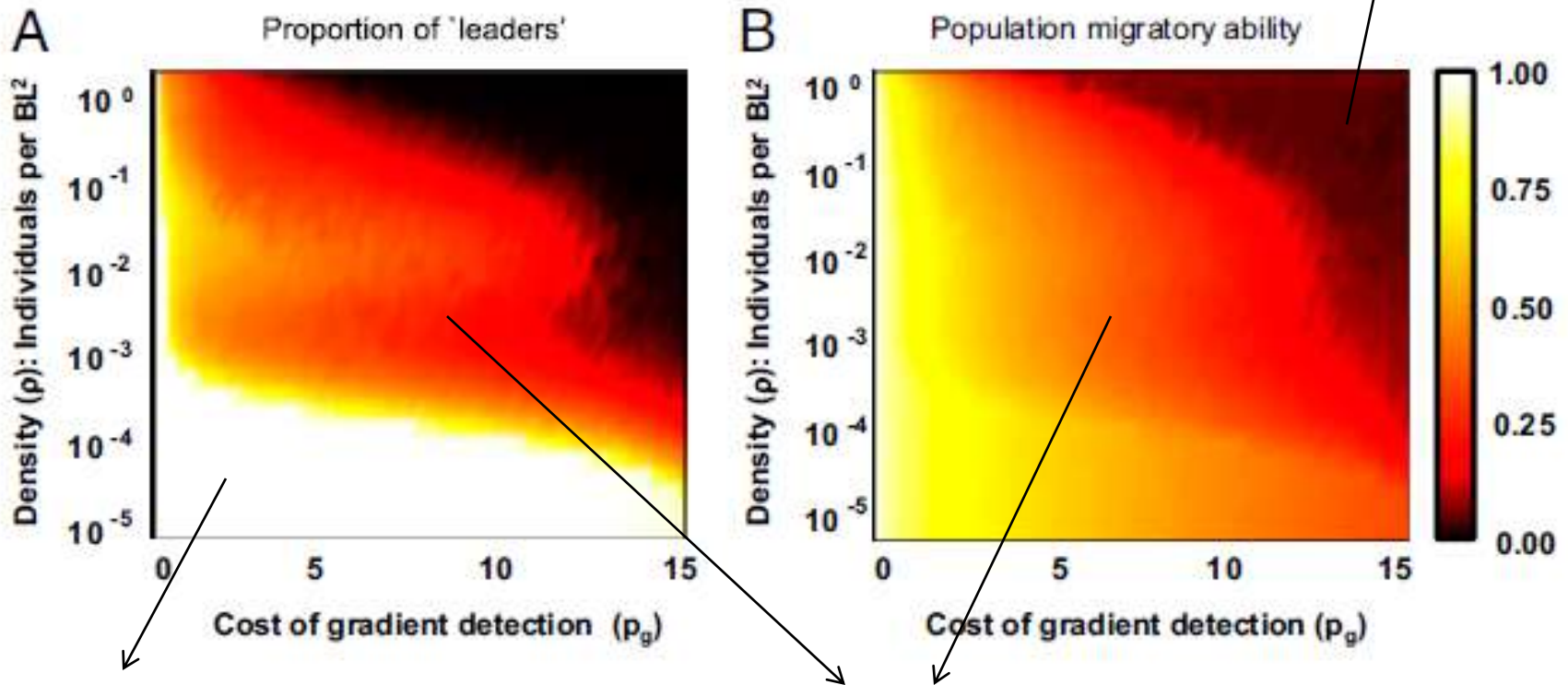


If we do not allow social Interactions:

i.e., for solitary individuals, the migration collapses at cost approx. 18 units.

Phase diagram of evolved populations

Very high density and/or high cost
=> No migration



Very low density and/or low cost.

=> Individual migratory strategies

Large intermediate region
Collective migratory strategies

Role of social interactions

- It allows *exploitation* of leaders *i.e.*, those who invest in sensing environment are exploited by *social individuals* who only follow others naively.
- They both coexist as a mixed strategy, resulting in collective migration.
- In our model, evolved collective migratory populations migrate less efficiently than solitary populations.

Acknowledgements

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