

# Swarm Intelligence

Eric Bonabeau

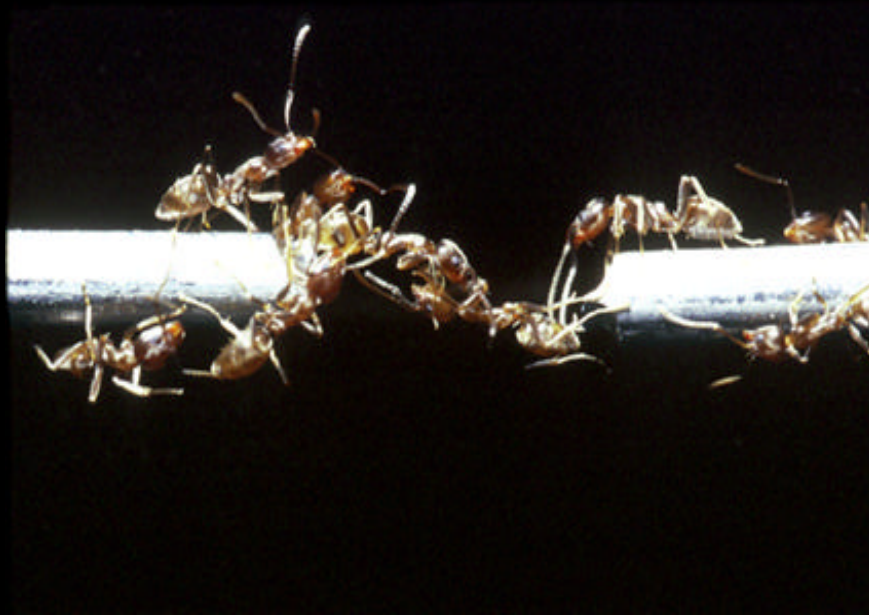
**i**cosystem

***Dumb parts, properly connected  
into a swarm, yield smart  
results.***

**Kevin Kelly**

**Great, but how do you properly  
connect the parts??**

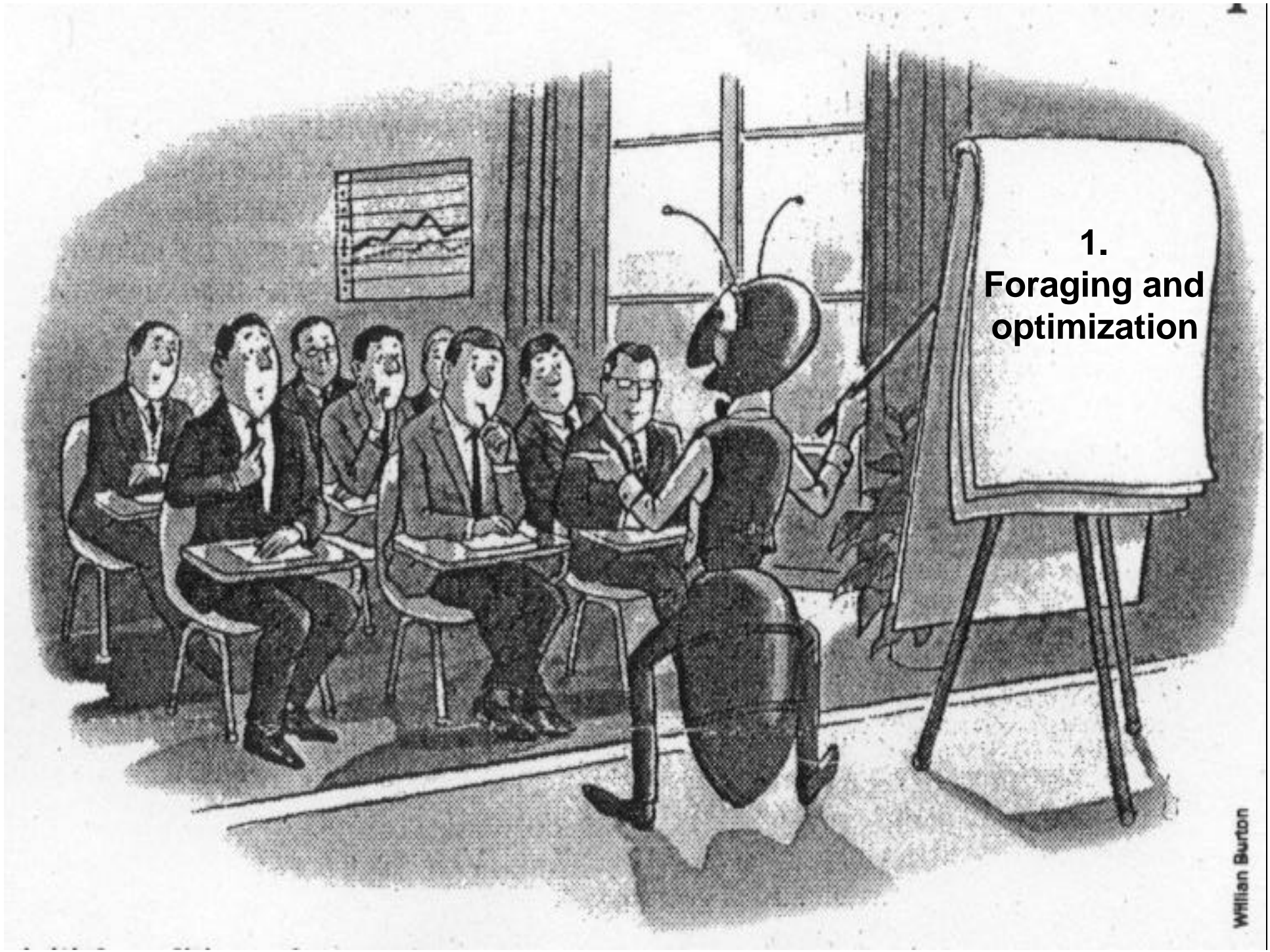
# Social insects do it

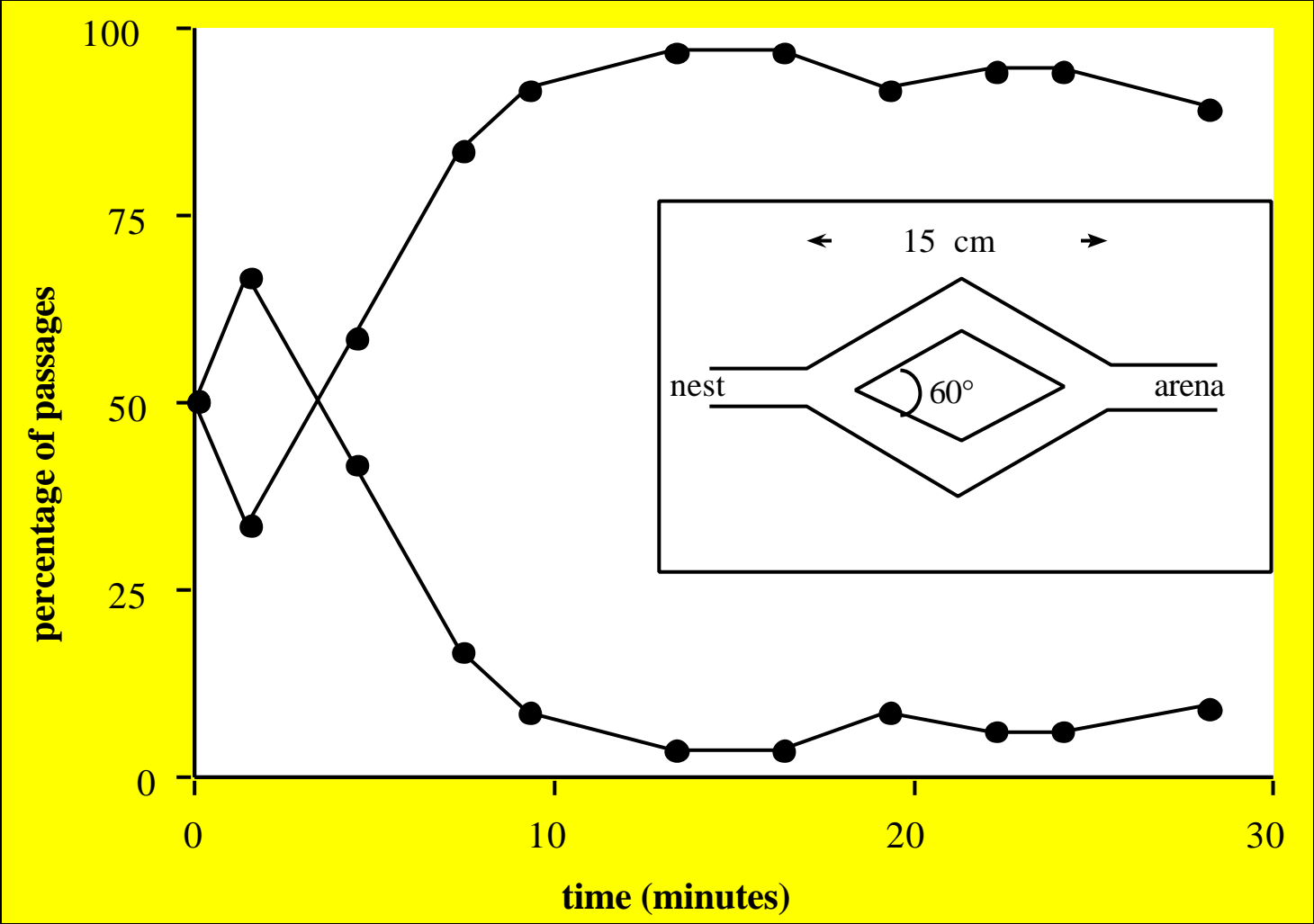


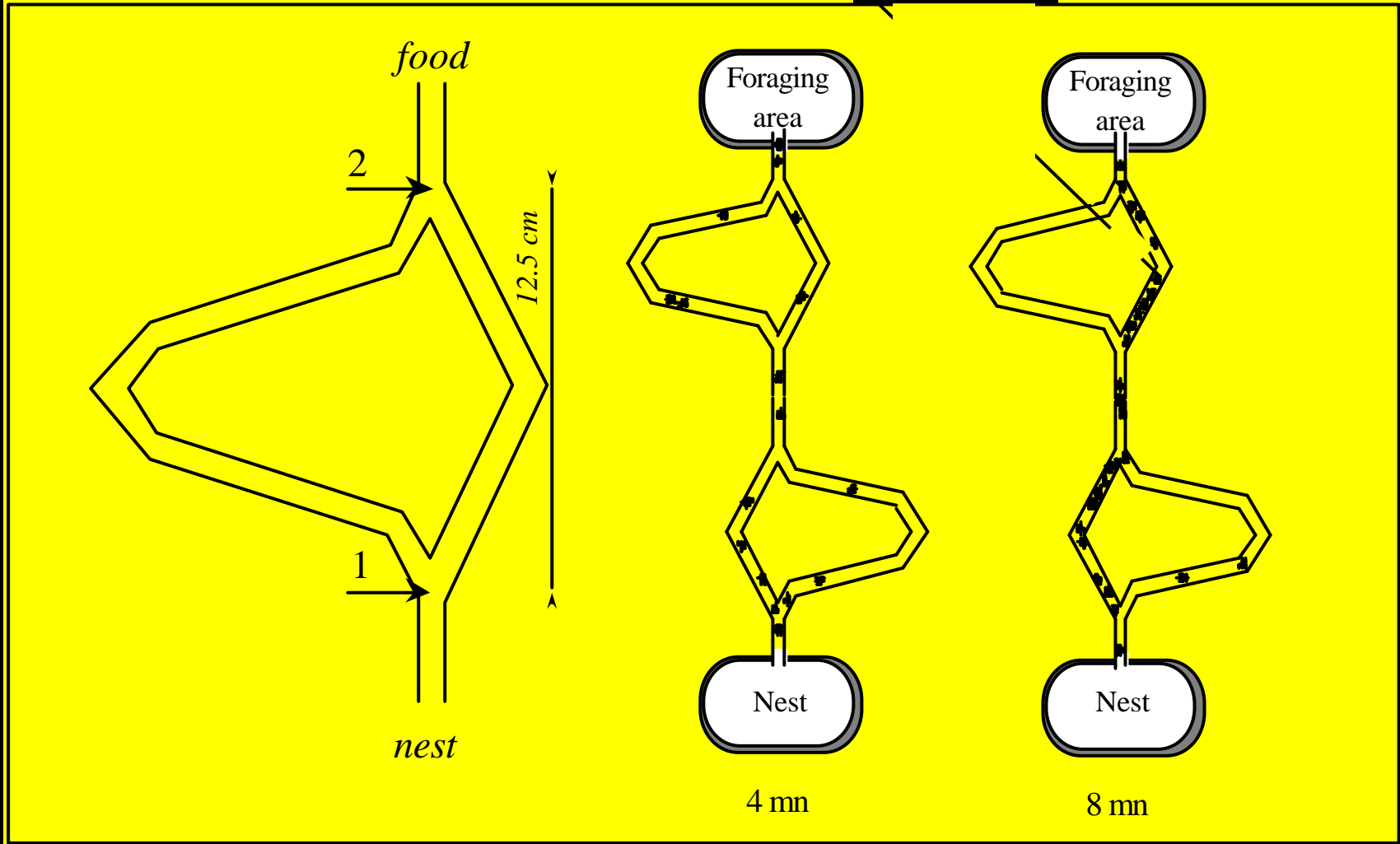
# A social insect colony is...

- ***Flexible***: the colony can respond to internal perturbations and external challenges
- ***Robust***: tasks are completed even if some individuals fail
- ***Decentralized***: there is no central control(ler) in the colony
- ***Self-organized***: paths to solutions are emergent rather than predefined

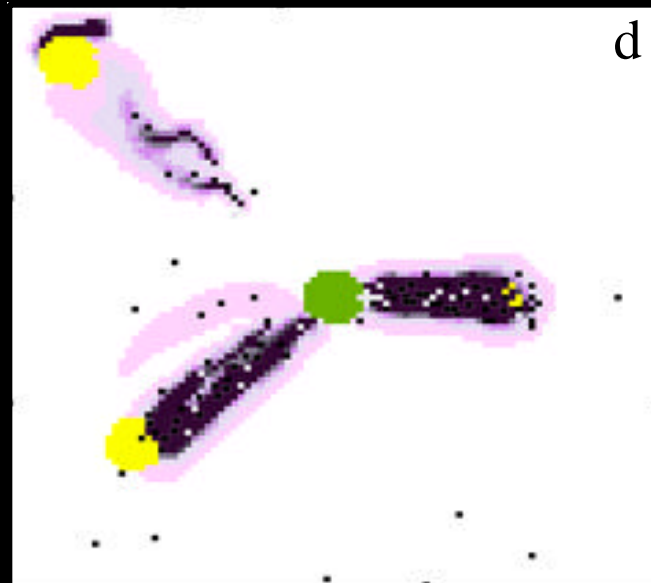
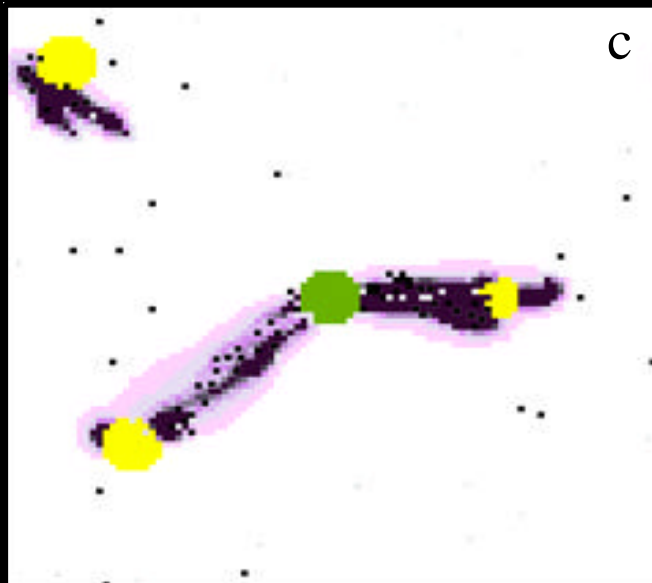
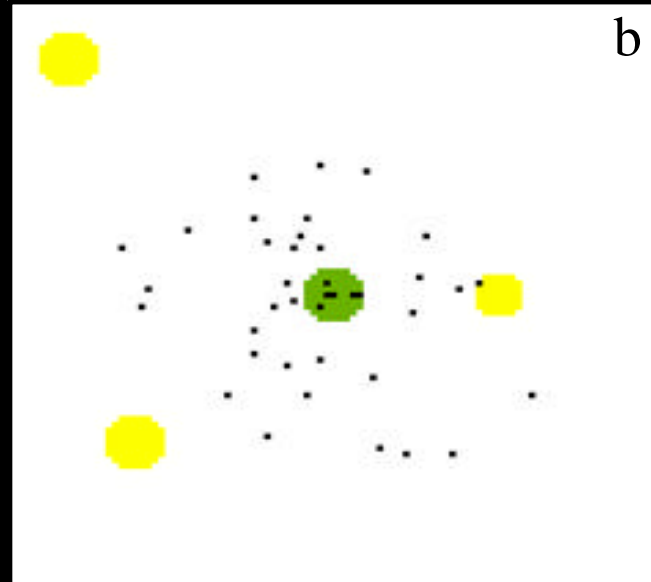
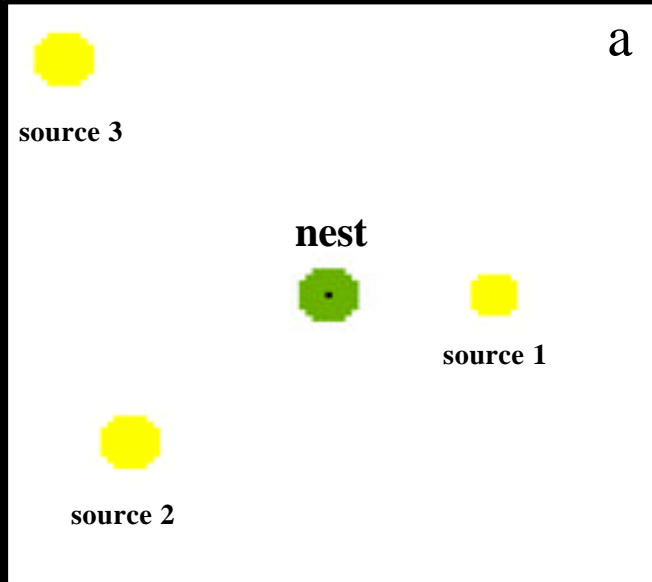
**1.  
Foraging and  
optimization**







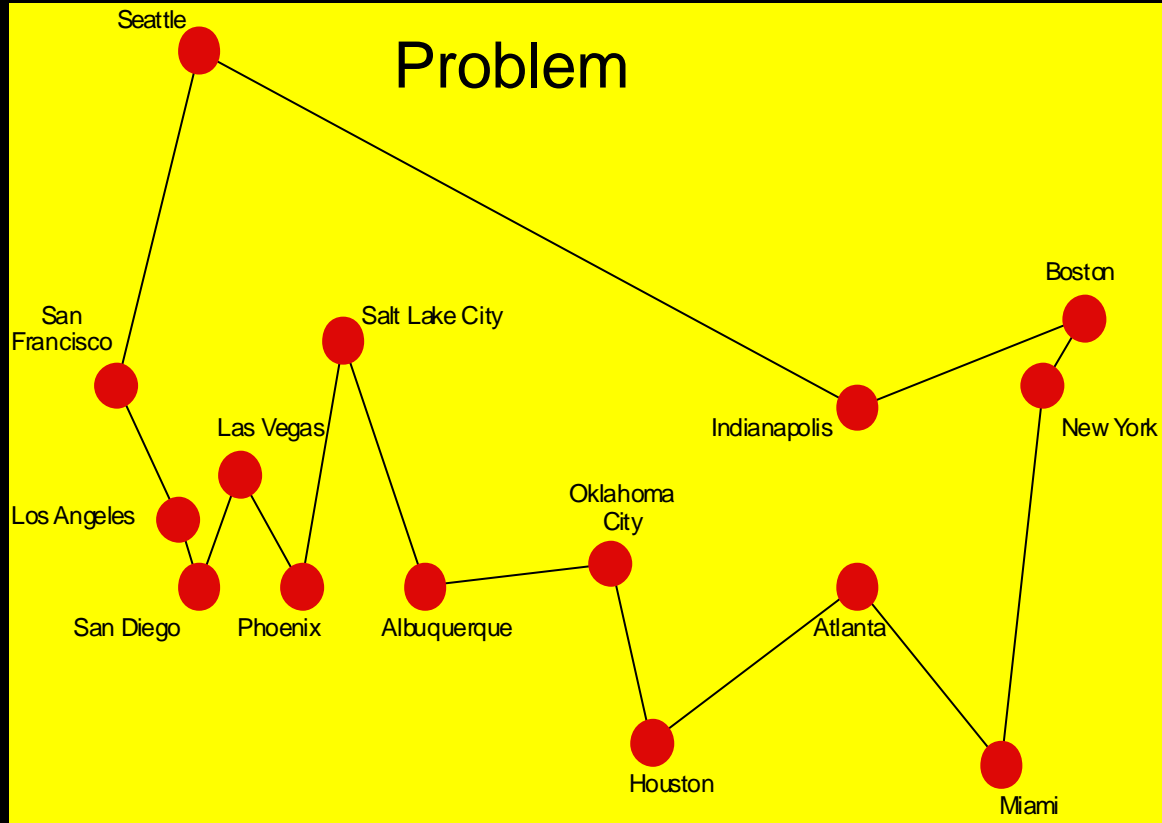




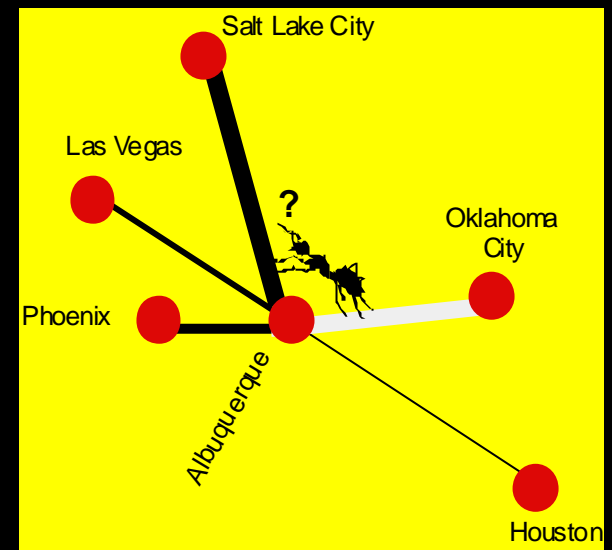
# Traveling sales-ants

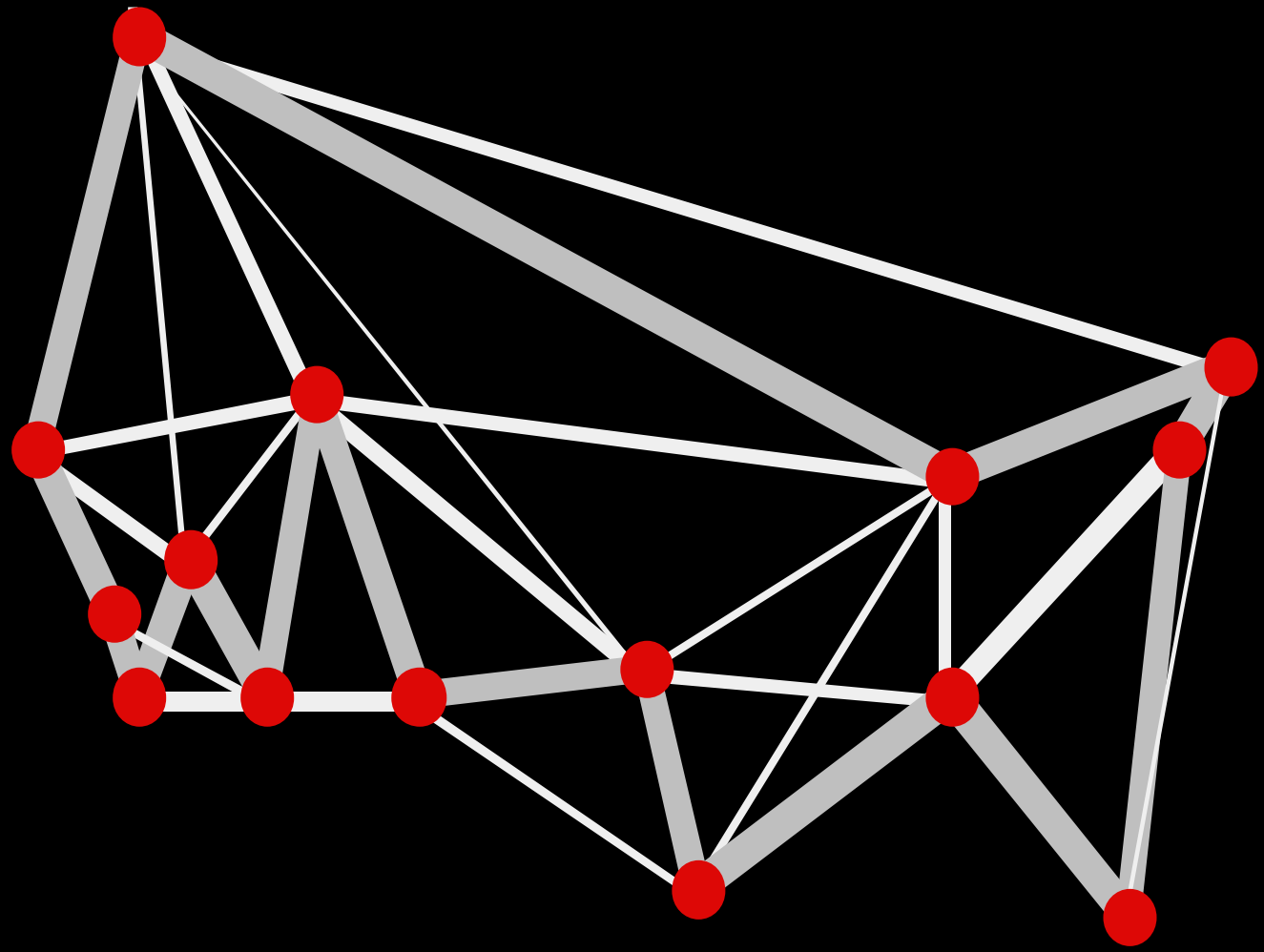
- $d_{ij}$  = distance between  $i$  and city  $j$
- $\tau_{ij}$  = virtual pheromone on link  $(i,j)$
- $m$  agents, each building a tour
- At each step of a tour, the probability to go from city  $i$  to city  $j$  is proportional to  $(\tau_{ij})^a (d_{ij})^{-b}$
- After building a tour of length  $L$ , each agent reinforces the edges it has used by an amount proportional to  $1/L$
- The virtual pheromone evaporates:  $\tau \longrightarrow (1-\rho) \tau$

# Problem



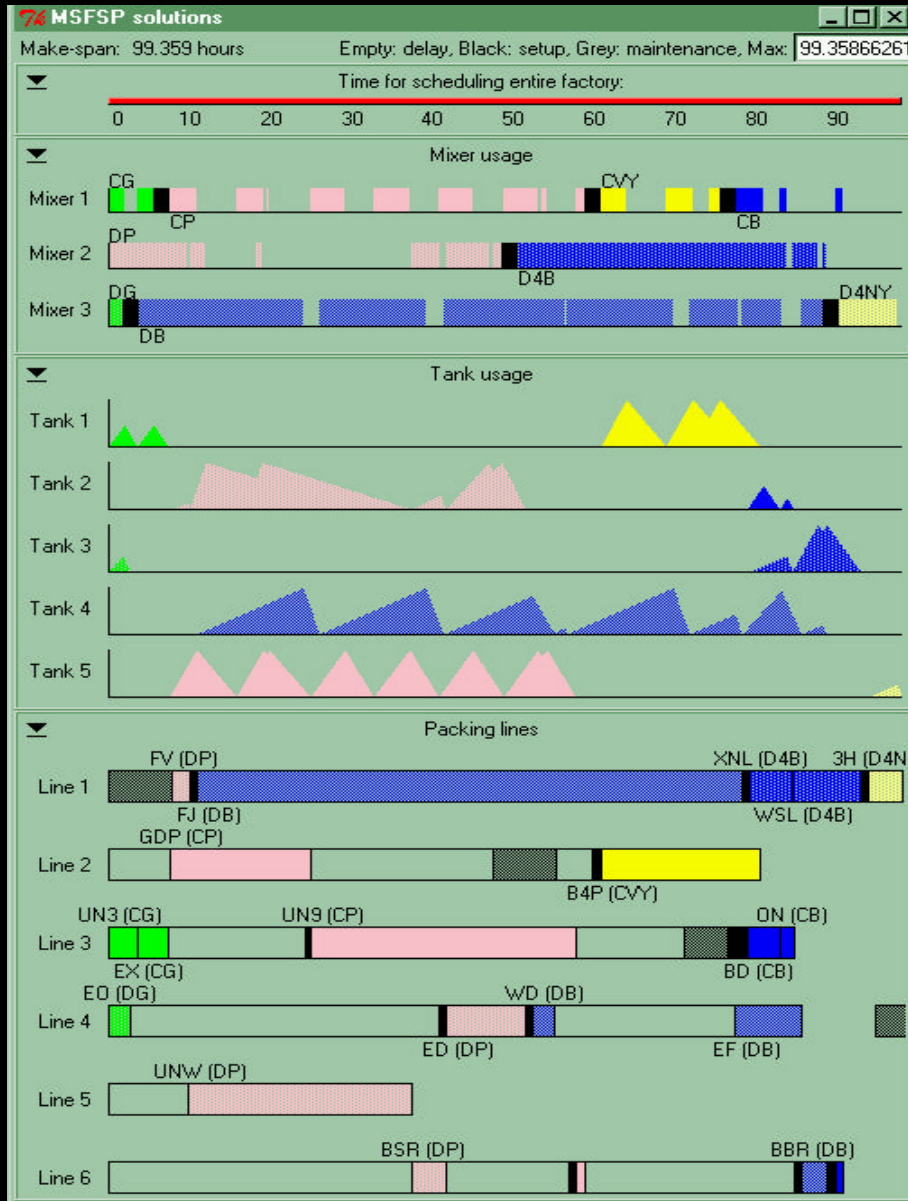
# The ant algorithm





# The same method applies to all assignment-type problems

- Traveling salesman problem
- Quadratic assignment problem
- Job-shop scheduling
- Graph coloring
- Vehicle scheduling



# Problem

- Production scheduling
- Dynamic problem with tough constraints

# Results

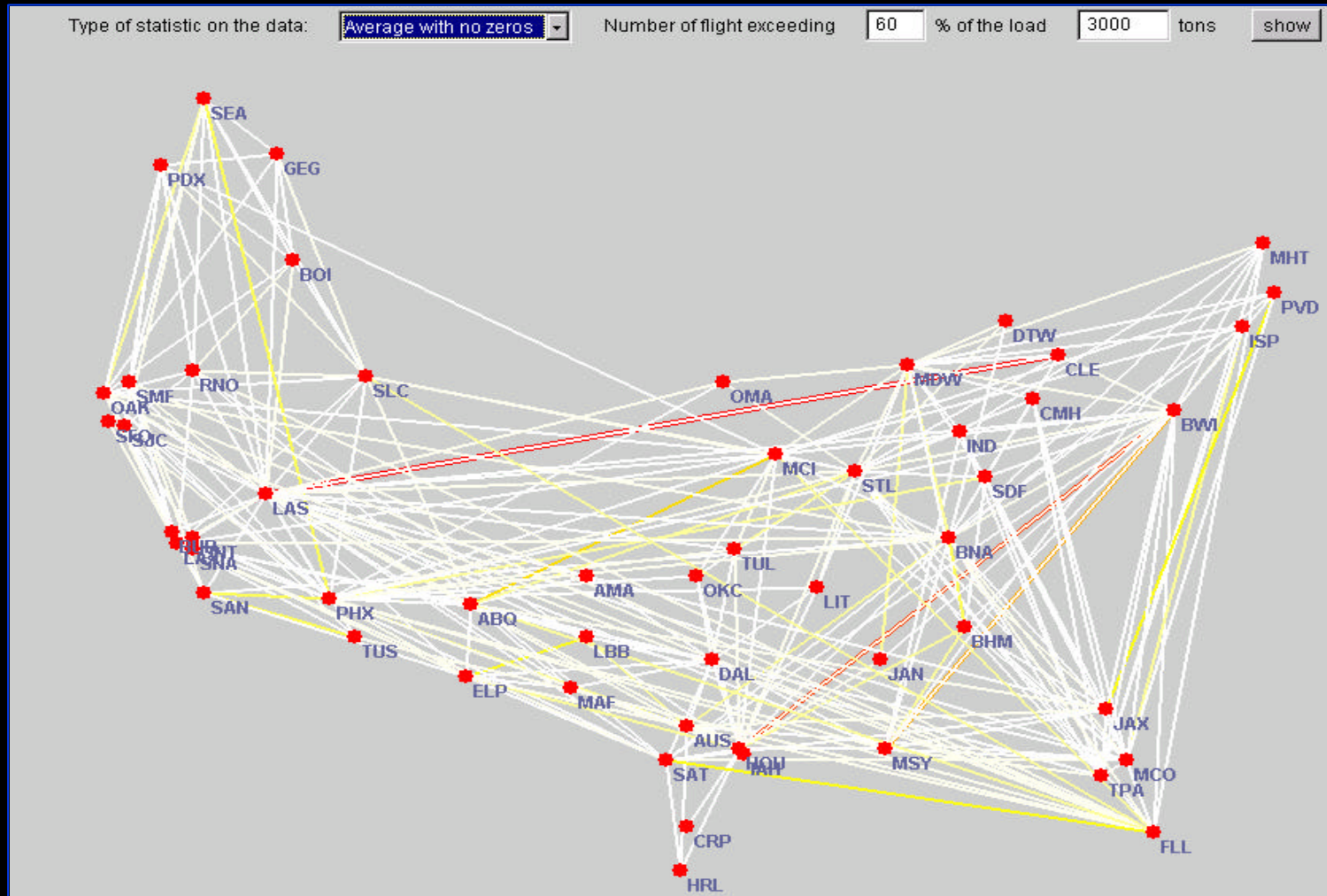
- Always finds a solution
- Better than best solution on market
- Order of magnitude faster!
- Copes with glitches and perturbations

# Problem

- Optimize routing
- Simple rules

# Results

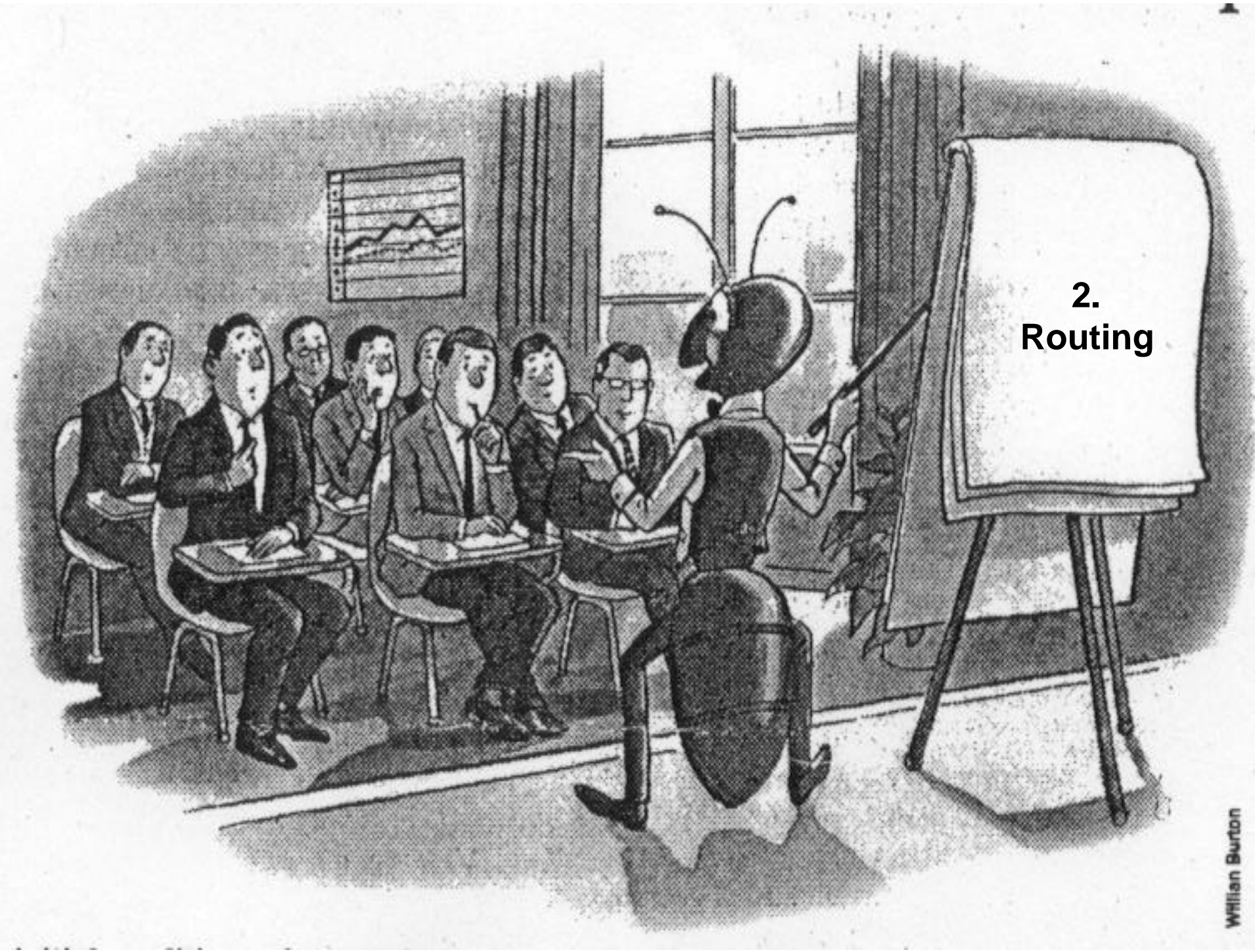
- 71% improvement
- At least \$10m/yr



## *Why does it work so well???*

- Local reinforcement of portions of solutions and global dissipation. Requires a certain problem structure (e.g., ant-based optimization is very good at solving structured [real-world] QAP instances).
- Keeping a distributed memory of exploration. Not only does this increase the efficiency of the optimization, it also facilitates its response to changing conditions because it can make use of previous knowledge about solution space.

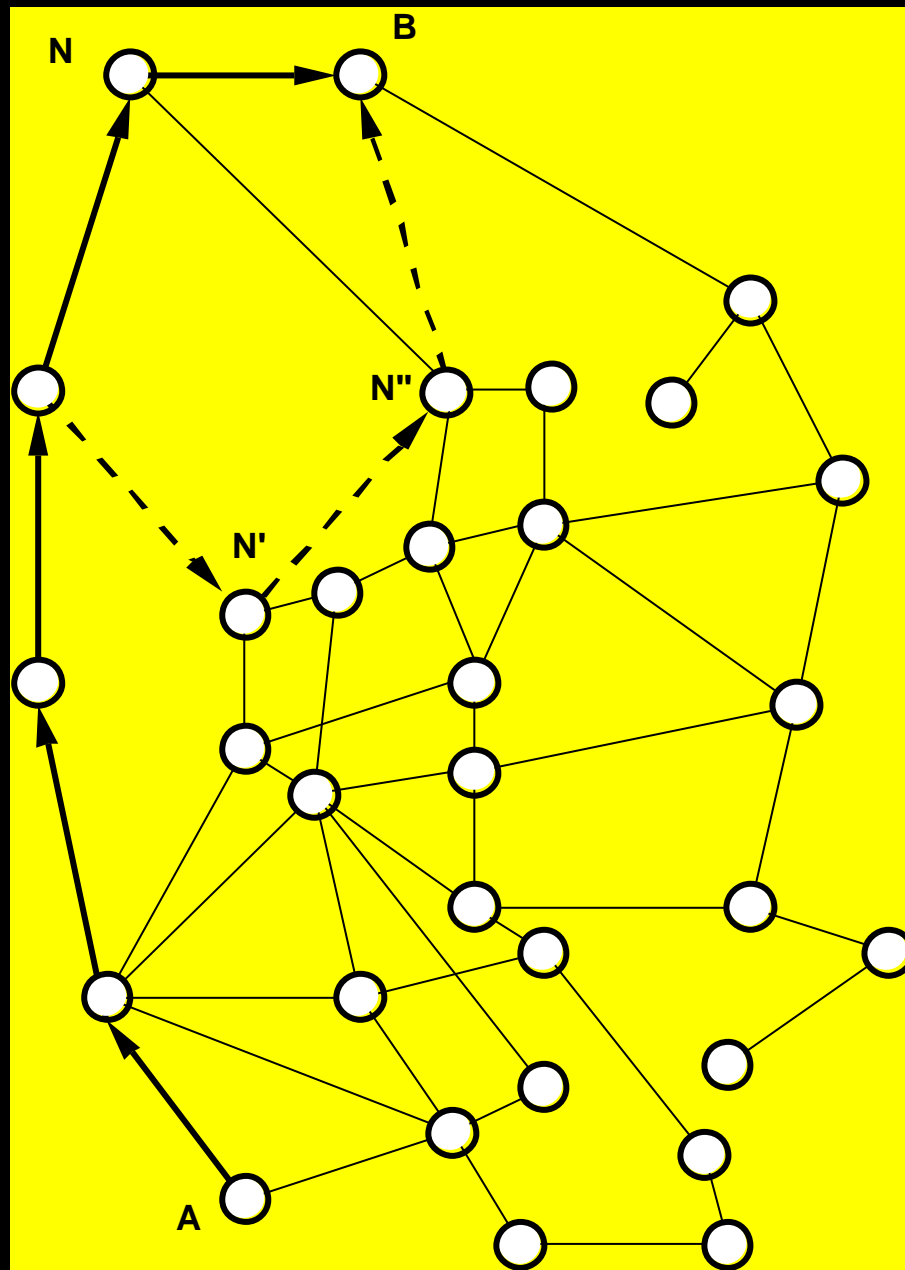




**2.  
Routing**

# Routing in communications networks

- Routing: mechanism that handles messages at switching stations.
- Messages must reach their destinations.
- It should take as little time as possible to go from source to destination.
- Traffic is constantly changing: routing must adapt.



# Routing

Switching stations or nodes have routing tables that tell messages where they should go next given their destination.

- Simple agents are launched in the network. Each agent goes from a source to a destination node.

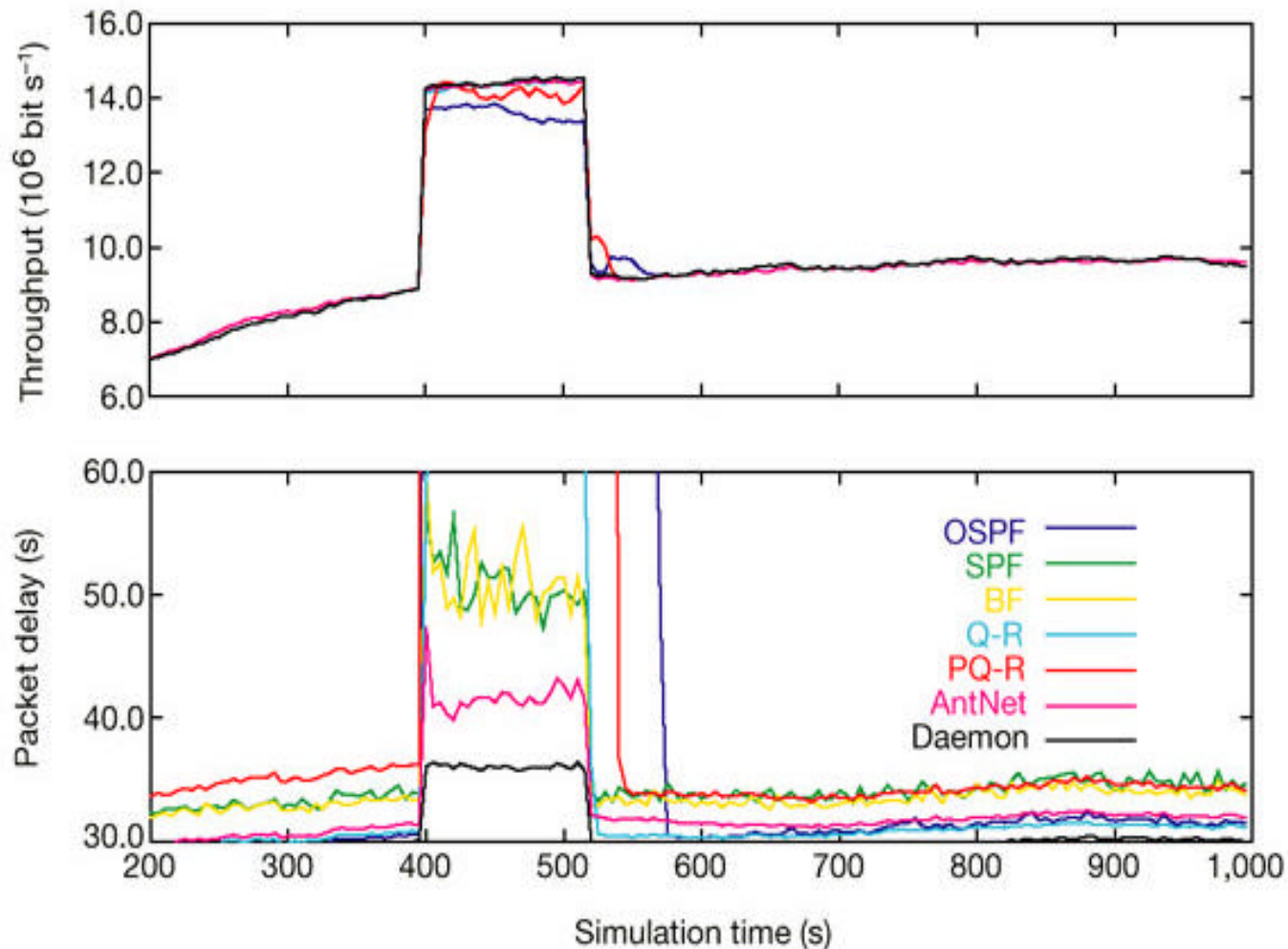
- An agent updates routing tables on its way to its destination, *viewing its source as a destination*.

"If you are going toward my source node, then hop to the node I am just coming from. Or don't."

- The influence of an agent decreases with its age.

- Agents are artificially delayed at congested nodes.

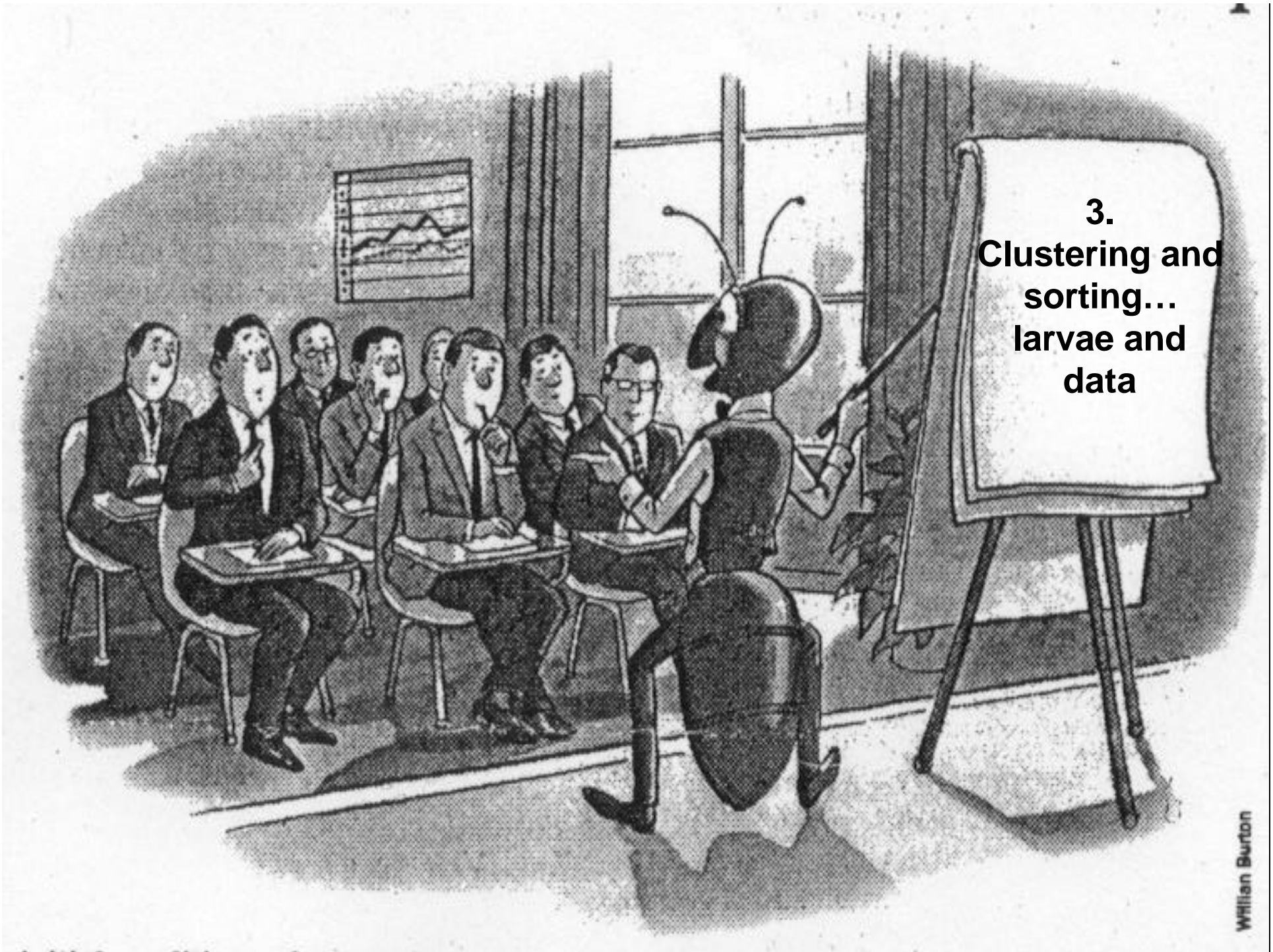
# Results (with AntNet)



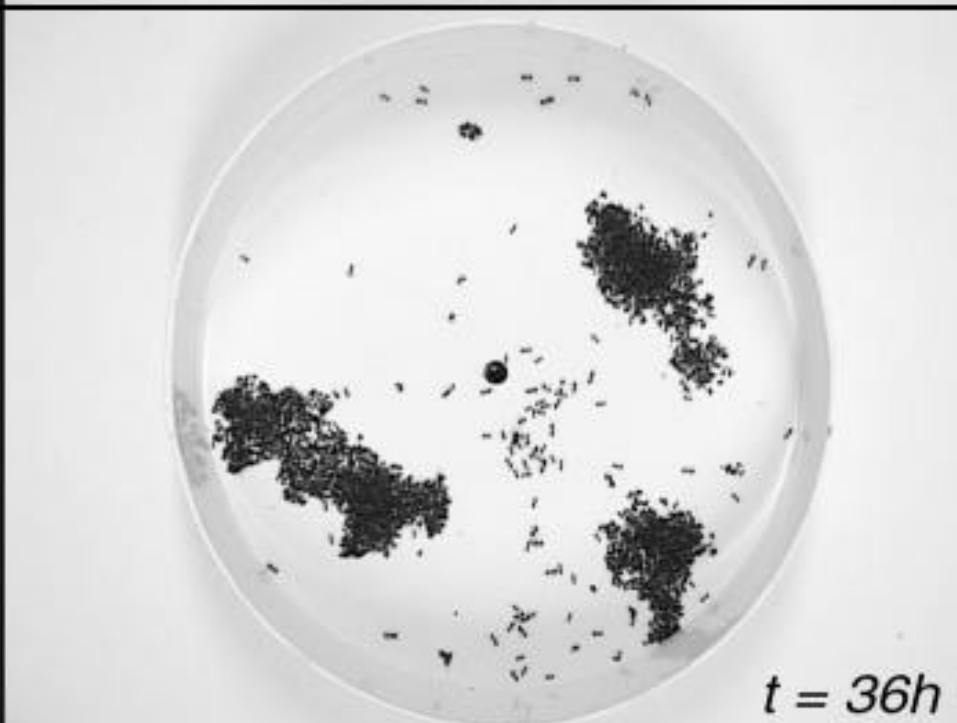
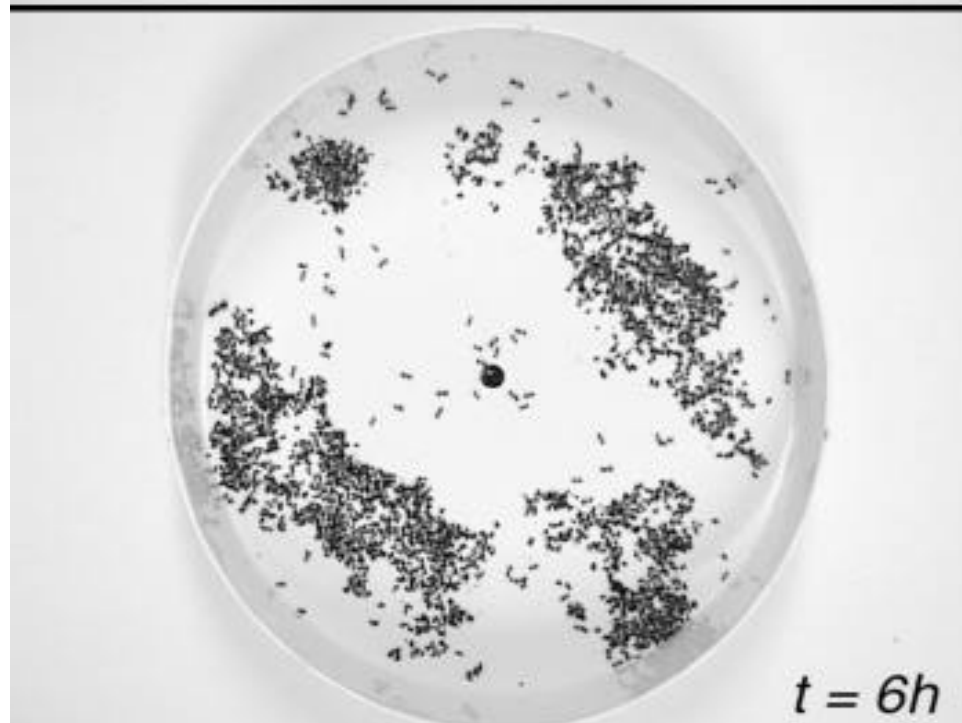
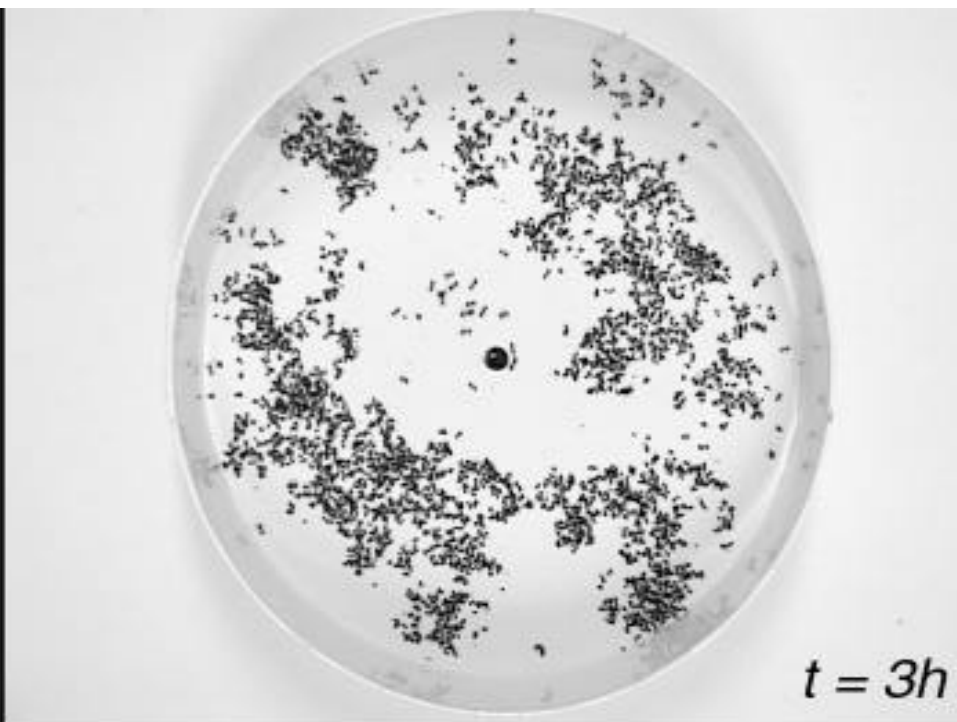
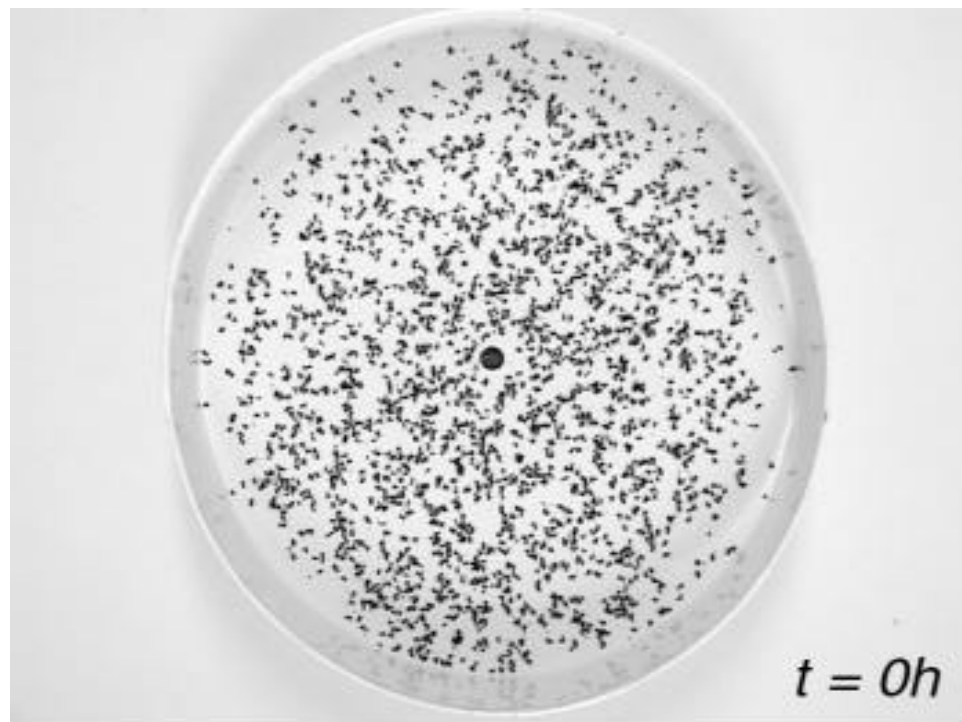
Hot spot  
superim-  
posed to  
Poisson  
traffic.

Moving  
average  
over 10s.

**3.  
Clustering and  
sorting...  
larvae and  
data**









# Clustering model

- An isolated item is more likely to be picked up by an unladen agent:

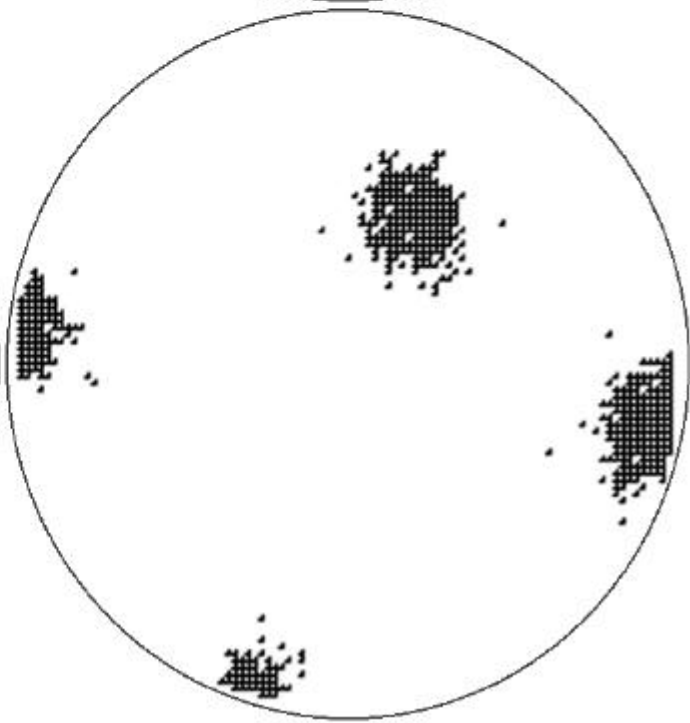
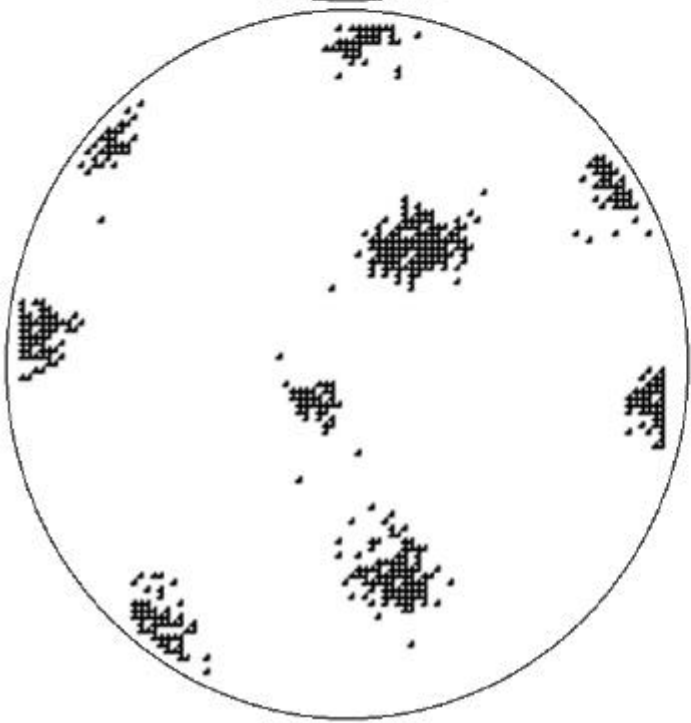
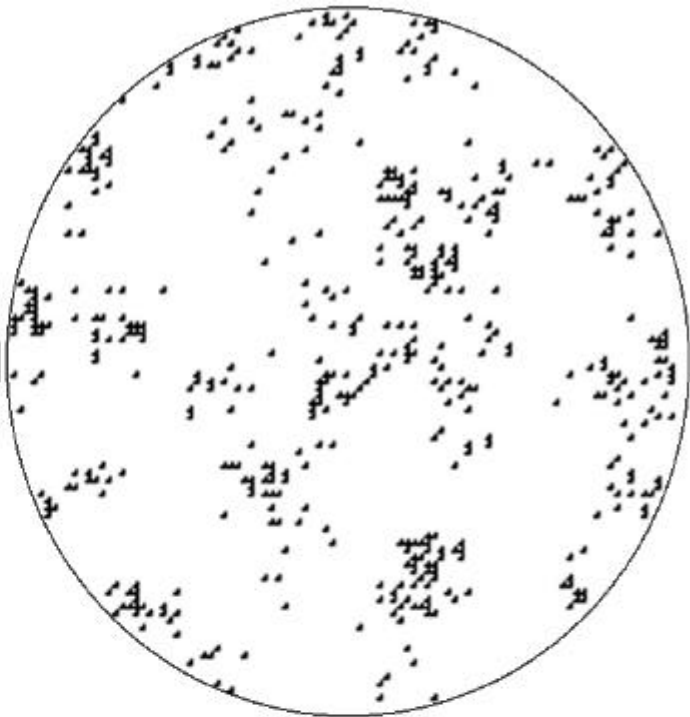
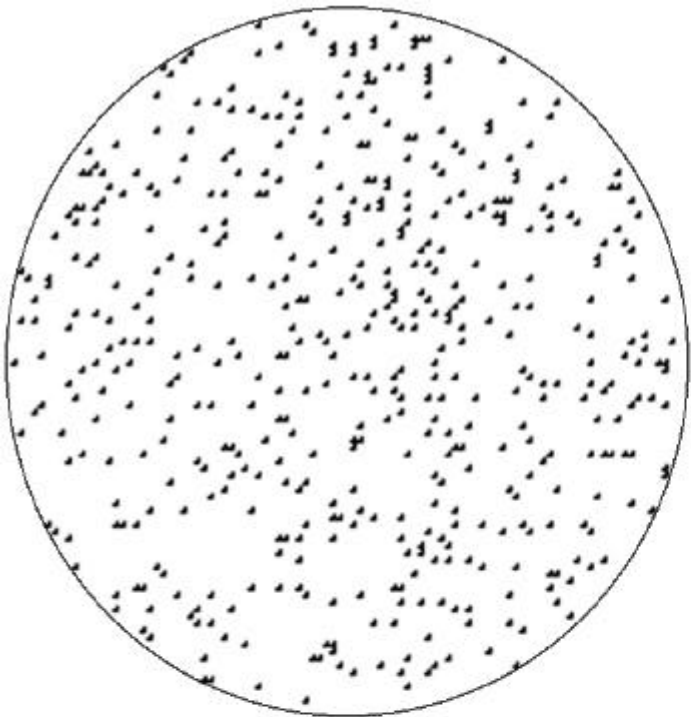
$$P_p = [k_1 / (k_1 + f)]^2$$

where  $f$  = density of items in neighborhood



- A laden agent is more likely to drop an item next to other items:

$$P_d = [f / (k_2 + f)]^2$$



# From clustering to sorting

- The same principle can be applied to sort items of several types ( $i=1, \dots, n$ ).

- $f$  is replaced by  $f_i$ , the fraction of type  $i$  items in the agent's neighborhood:

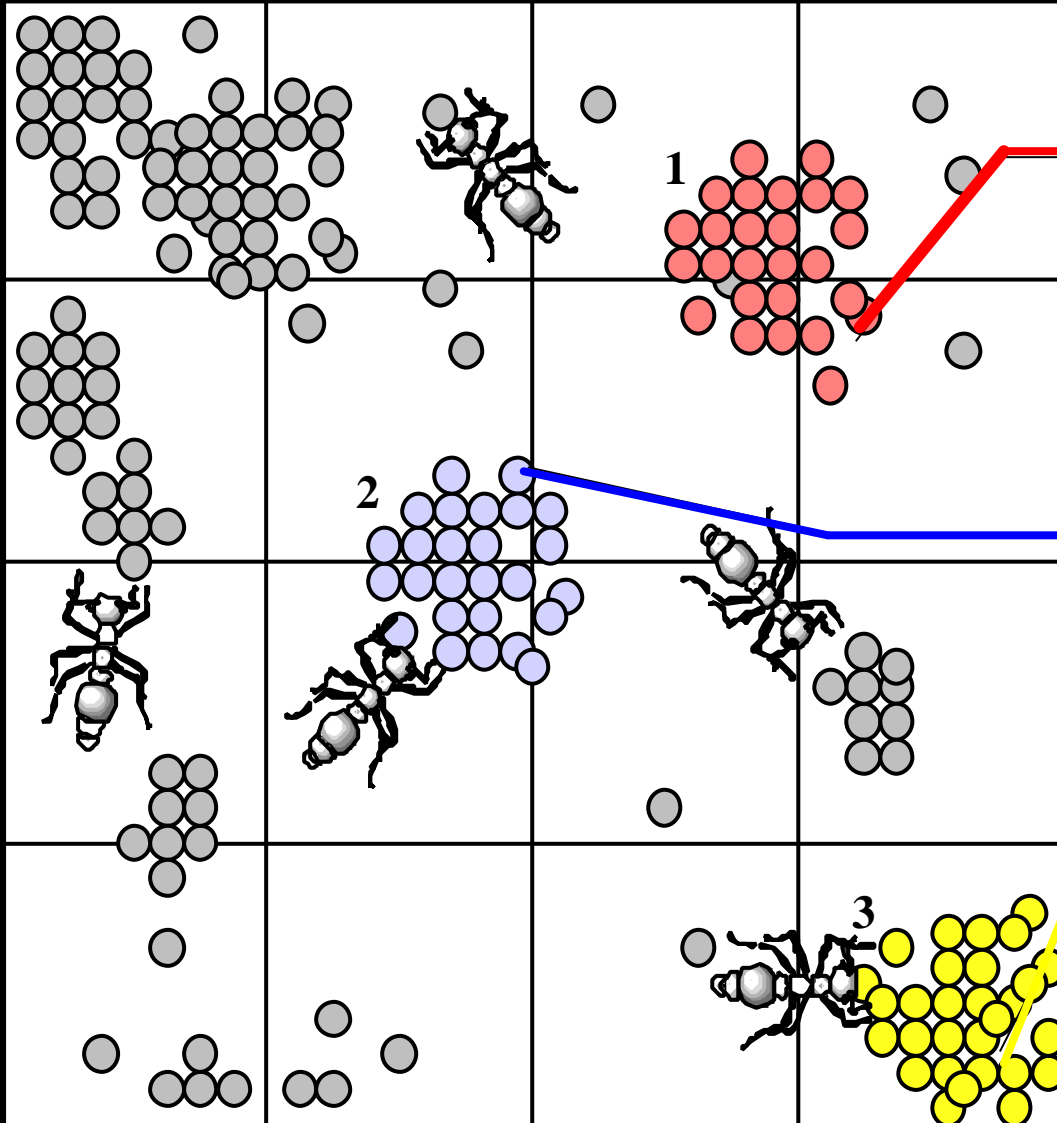
$$P_p(i) = [k_1 / (k_1 + f_i)]^2$$

$$P_d(i) = [f / (k_2 + f_i)]^2$$



If items are described by real-valued attributes (points in  $\mathbb{R}^n$ ), the same principle can still be applied:  $f$  is now replaced by a normalized distance between the item carried by the agent and items in the agent's neighborhood.

 Items will end up being next to items with close attributes.

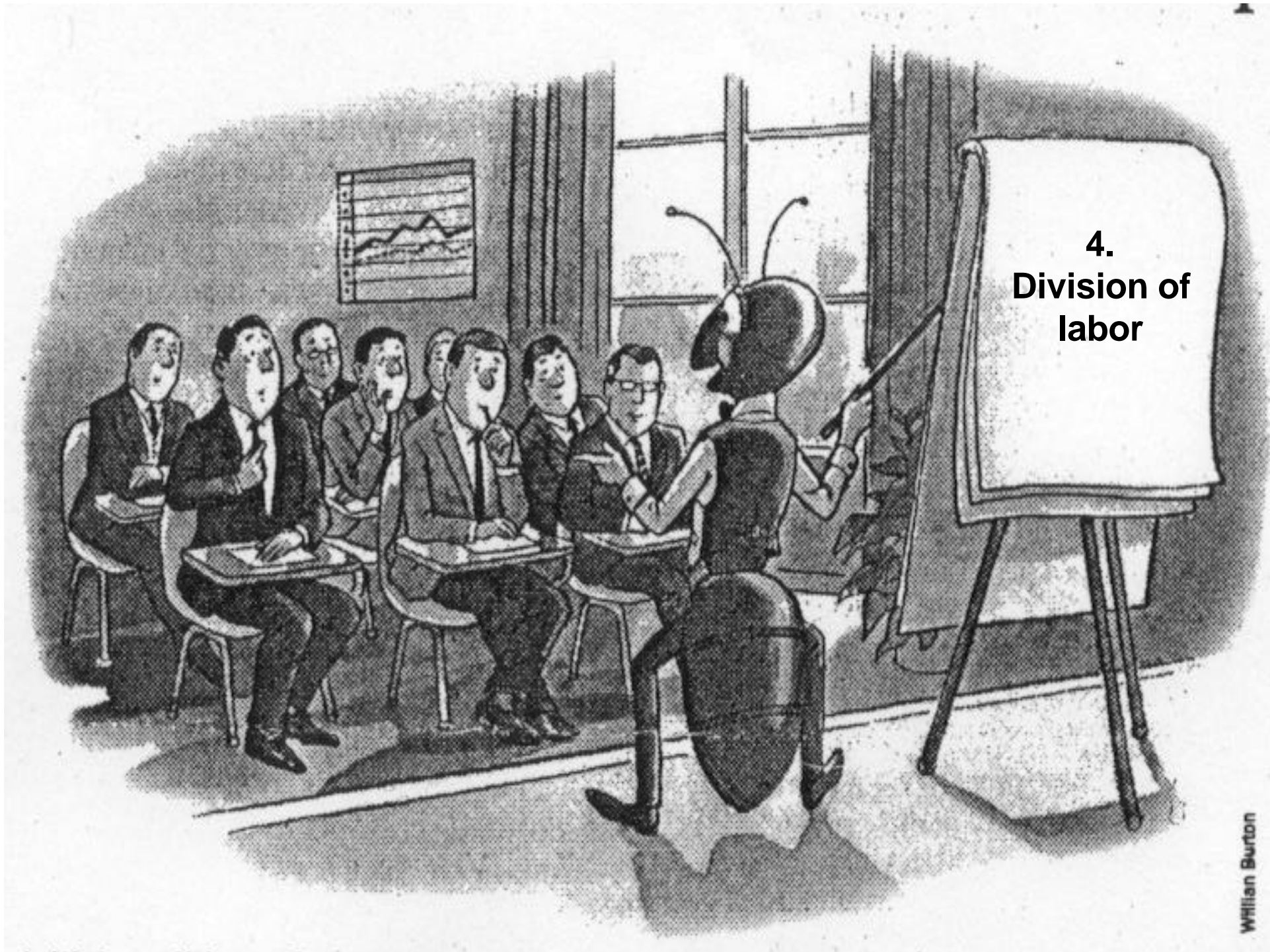


Single  
Male  
Age: 20  
Banking product: interest checking  
lives with parents

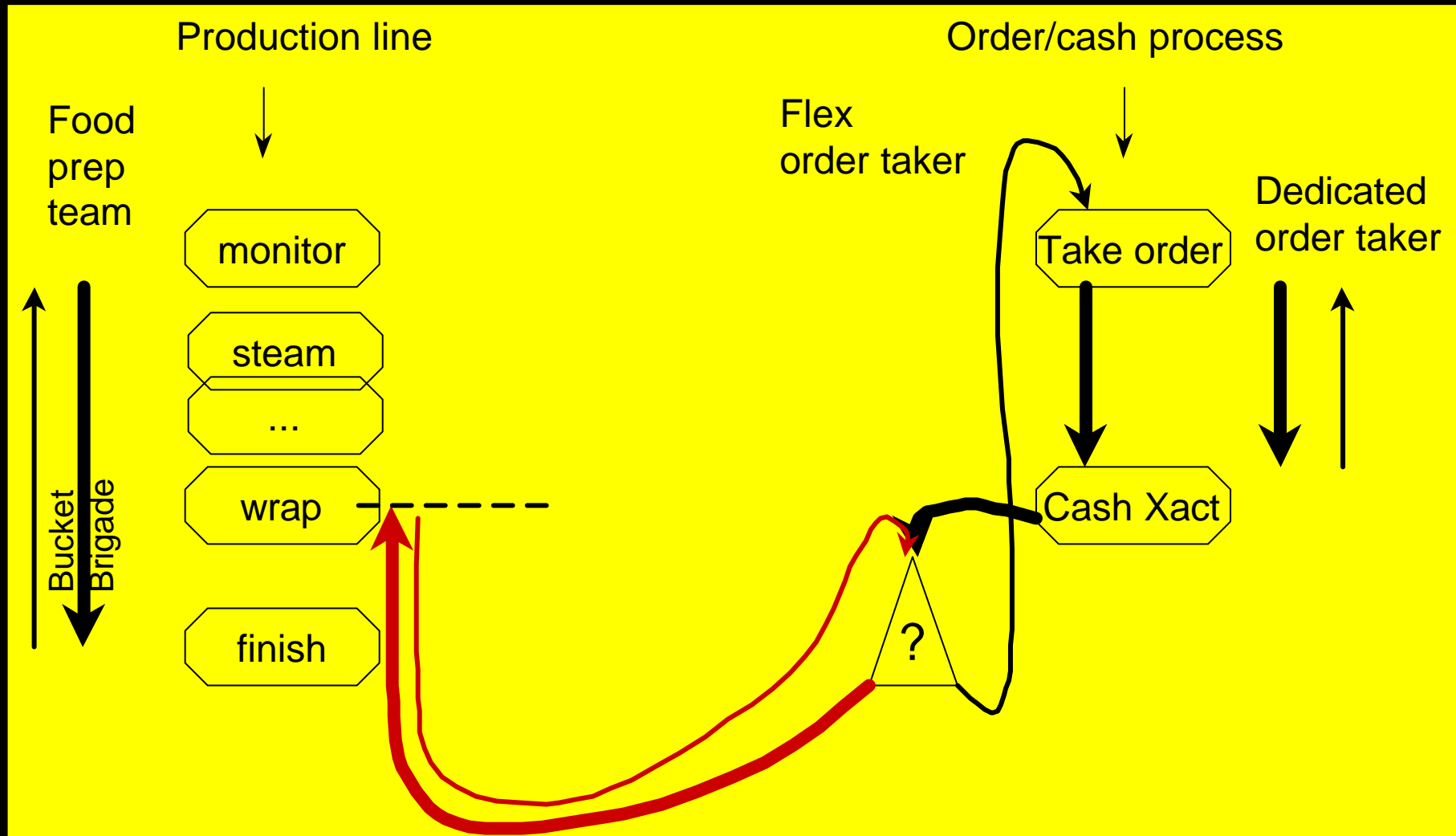
Married  
Female  
Age: 57  
No mortgage  
Home owner

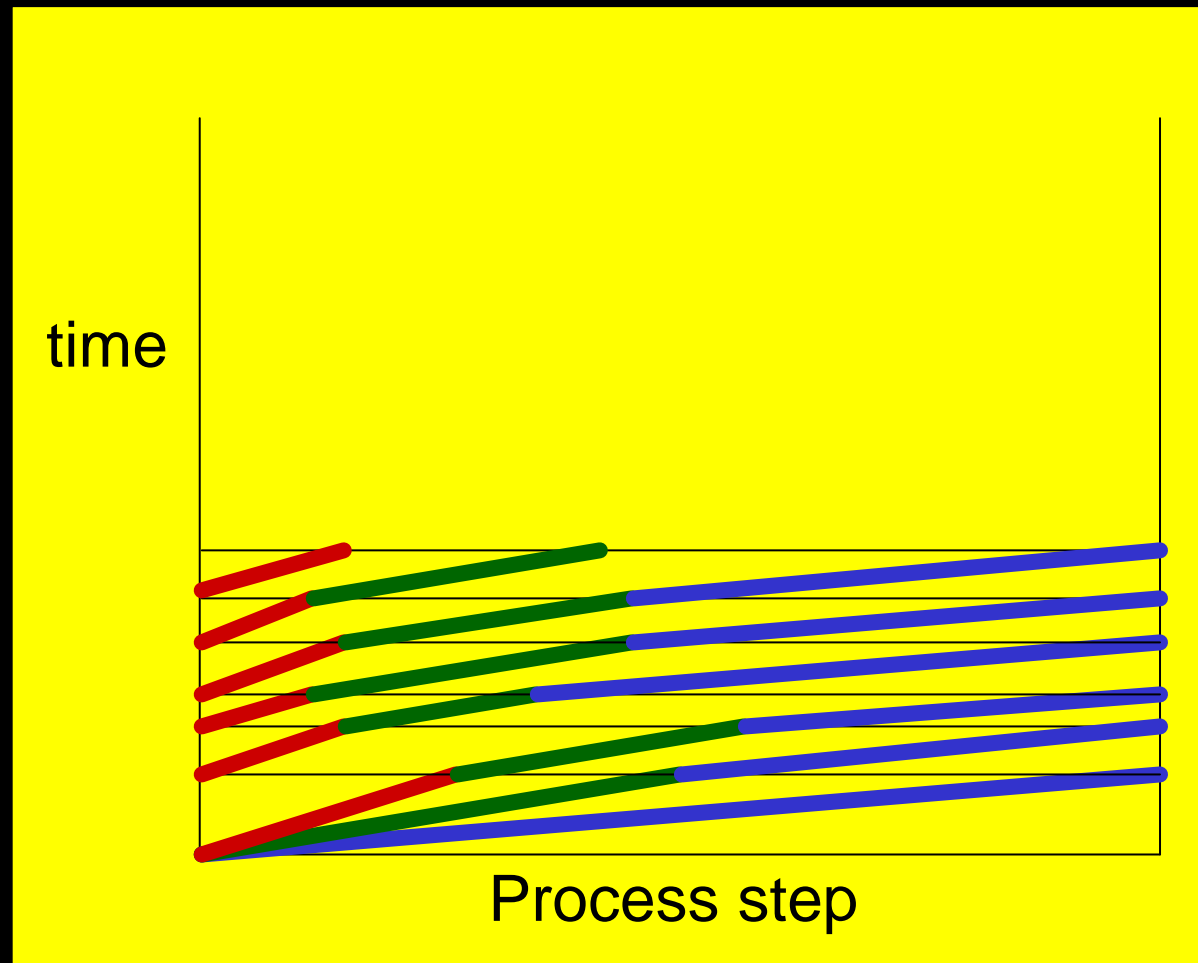
Married  
Male  
Age: 44  
Banking product: savings  
Tenant

**4.  
Division of  
labor**

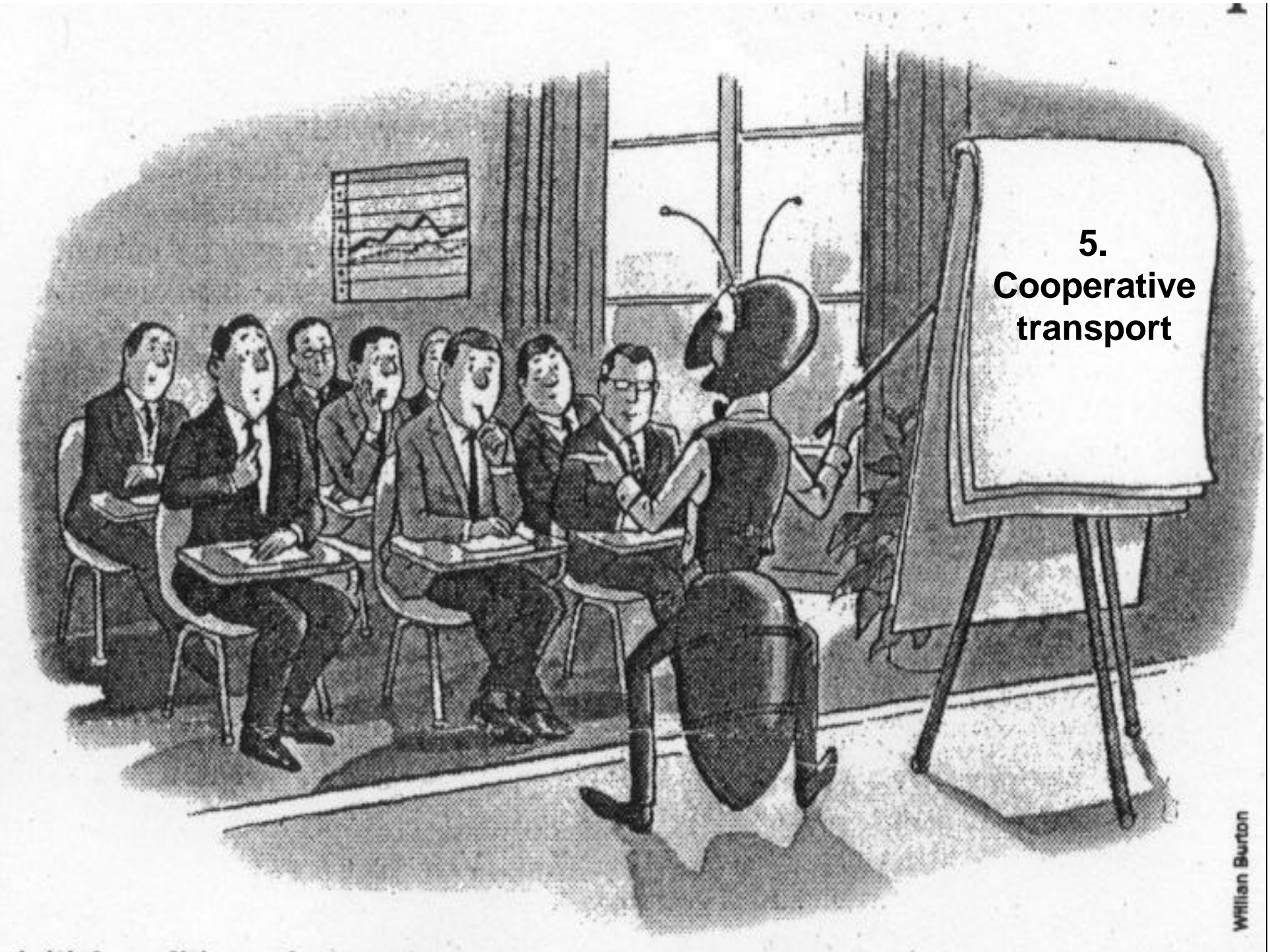


# Bucket brigades at Taco Bell



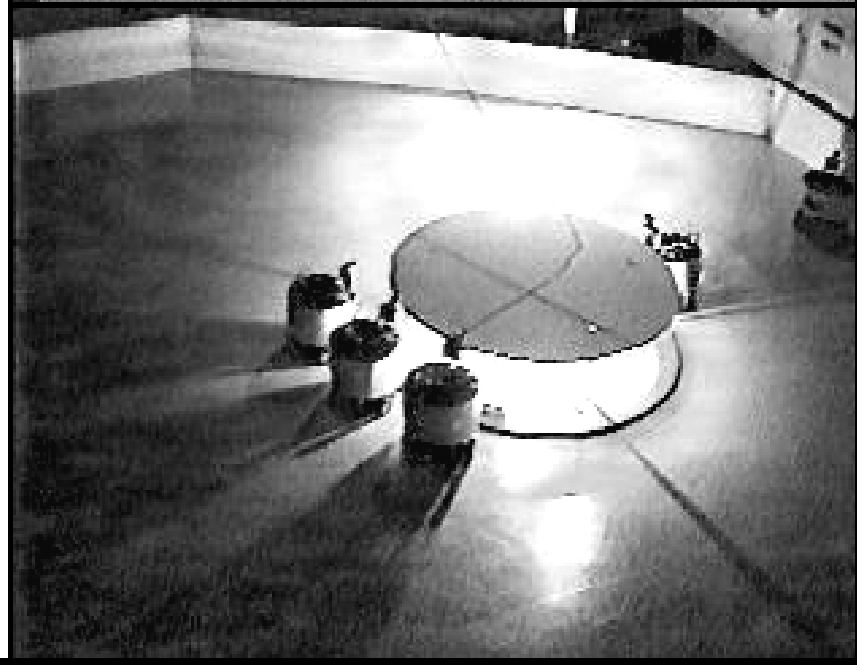
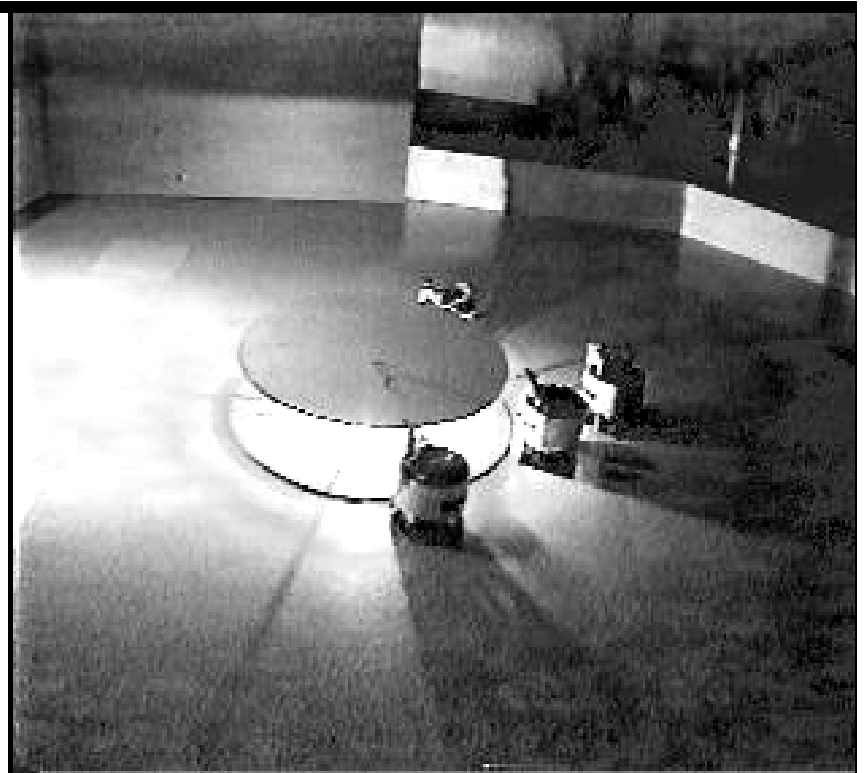
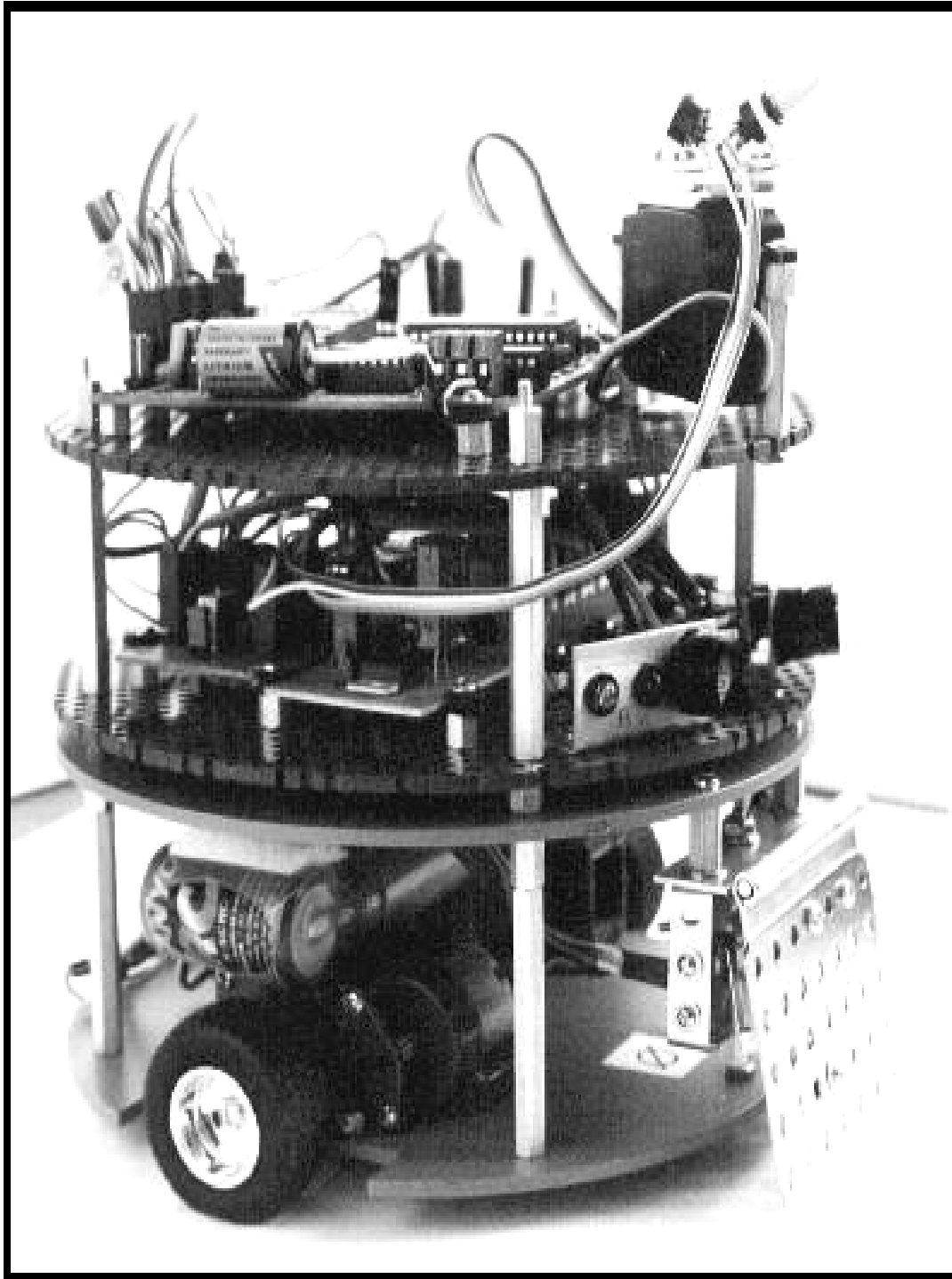


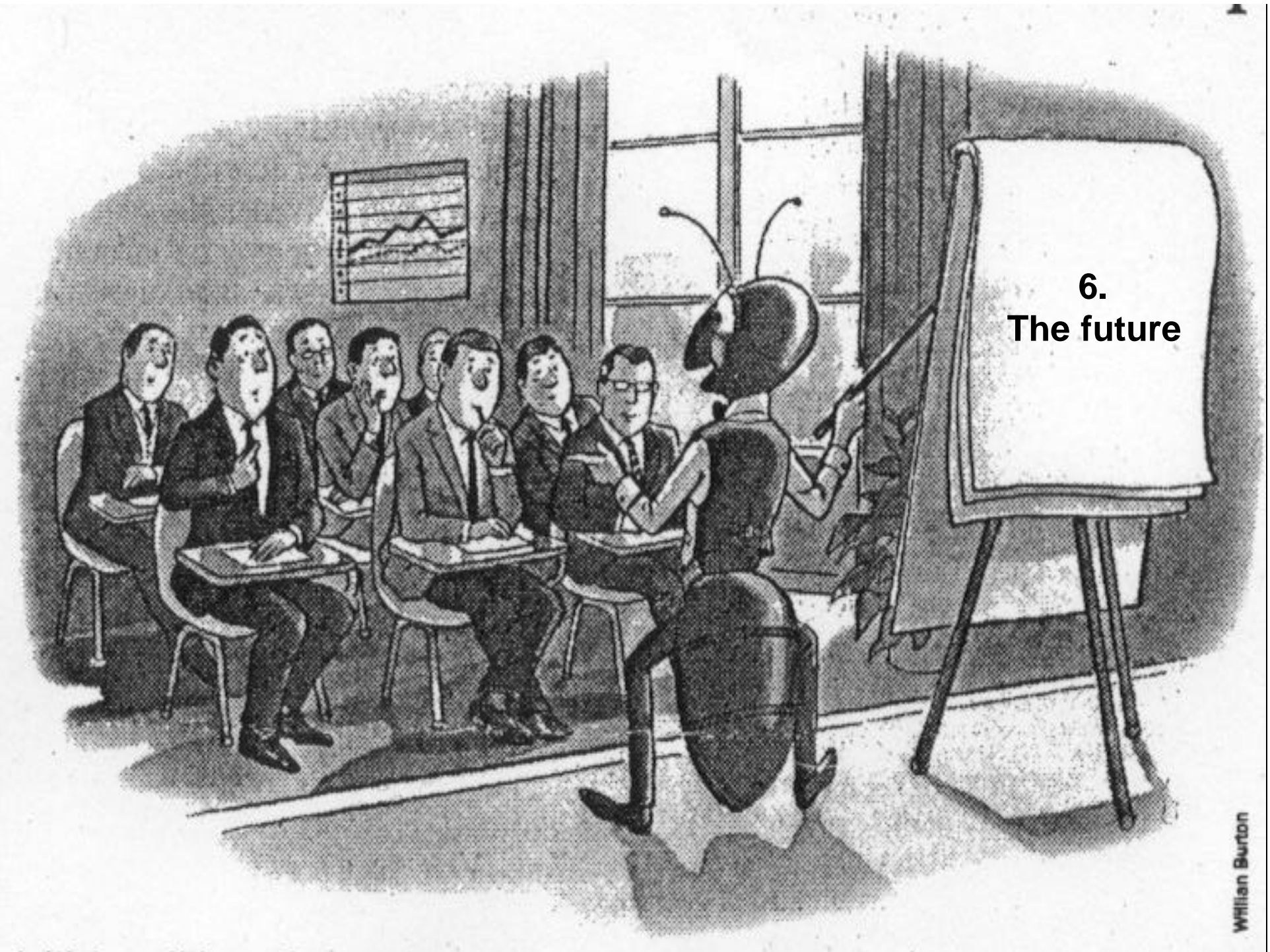




**5.  
Cooperative  
transport**







**6.**  
**The future**

Miniaturization

Satellite

Maintenance

Pipe Inspection

Cleaning Ship

Hulls

Engine

Maintenance

Medical

Self-Assembling  
Robots

Telecommunications

Job Scheduling

Combinatorial  
Optimization

Interacting Chips  
in Mundane  
Objects

Pest Eradication

Vehicle Routing

Optimal  
Resource  
Allocation

Data Clustering

Distributed Mail  
Systems