

# Models of Language Evolution

Agent-Based Models

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# Goals for today

- 1 look at 3 case studies of agent-based models for meaning evolution
  - 1 cellular automata
  - 2 naming game
  - 3 category game
- 2 see what's good and bad about each of these

# Conway's Game of Life

- grid of cells
- each cell  $x$ :
  - has 8 neighbors
  - is alive or dead at any given time
- simultaneously update all cells:
  - 1 any live cell stays alive iff it has exactly 2 or exactly 3 live neighbors
  - 2 any dead cell becomes alive iff it has exactly 3 neighbors

## Meaning Evolution in Cellular Automata

- finite grid of agents, with 8 neighbors each
- there are randomly walking *predators* and *food sources*
- each round each agent has a choice whether or not to do any of the following (coded as a bitvector  $\vec{x} \in \{0, 1\}^4$ ):
  - (i) open mouth
  - (ii) hide
  - (iii) emit sound 1
  - (iv) emit sound 2
- each action incurs some (non-positive) cost  $\vec{c} \in \mathbb{R}^4$
- agents receive positive payoffs  $f$  for opening the mouth when in a cloud of food
- agents receive negative payoffs  $b$  when *not* hiding in a cloud of predators

## Meaning Evolution in Cellular Automata

- each agent  $i$  can condition her choice on whether or not any of the following happen in the previous round (coded as a bitvector  $\vec{y}$ ):

(i)  $i$ 's been fed

(iii)  $i$  heard sound 1

(ii)  $i$ 's been hurt

(iv)  $i$  heard sound 2

- agents have 265 strategies in total (all functions from  $\vec{y}$  to  $\vec{x}$ )
- each agent  $i$  gets a **reward** for each round  $t$  depending on his actions  $\vec{x}$ :

$$R(i, t) = b + f + \vec{x} \cdot \vec{c}$$

- we consider the **accumulated rewards** (ARs) between round  $t$  and  $t'$ :

$$AR(i, t, t') = \sum_{t \leq \tau \leq t'} R(i, \tau)$$

- every 100 rounds each agent compares the ARs of her neighbors and adopts the *strategy* of the most successful neighbor (“**imitate-the-best dynamics**”)

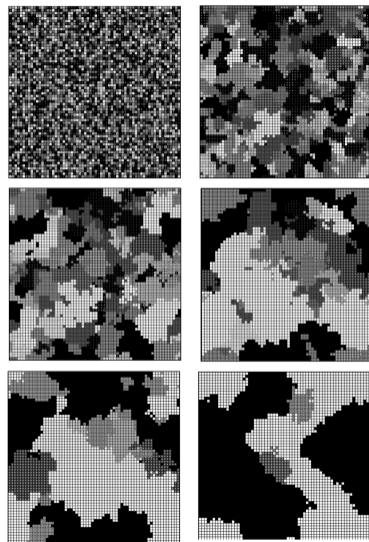
→ What's going to happen? ←

## Result of Simulation

- starting from a random population
- regions of “perfect communicators” emerge:
  - perfect communicators use one signal for food, one for predators

## Reflection

- is this a good / plausible model of meaning evolution?
- anything we would like to know further about the model?



## (Minimal) Naming Game

- population of  $n$  agents looking for word for one object/meaning
- at each point in time each agent has a vocabulary of words
- initially all agents have one random word
- asynchronous update with actual play

# (Minimal) Naming Game: Play & Update Rule

let

$V_S$ : vocabulary of speaker

$V_H$ : vocabulary of hearer

select  $w \in V_S$  uniformly at random

if  $w \in V_H$ : (play is a success)

$V_S \leftarrow \{w\}$

$V_H \leftarrow \{w\}$

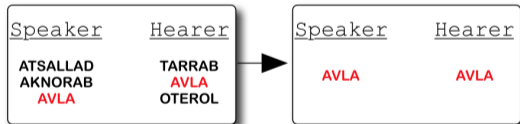
otherwise: (play is a failure)

$V_H \leftarrow V_H \cup \{w\}$

**Failure**

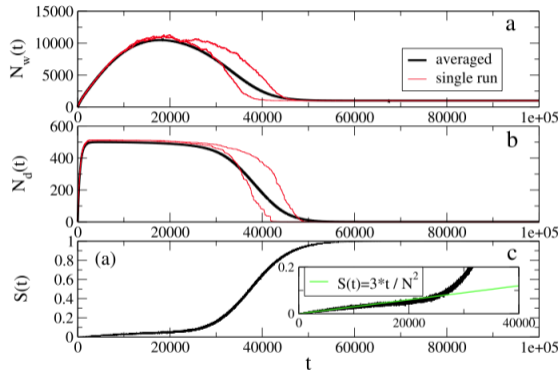


**Success**





## (Minimal) Naming Game: Results



**Fig. 2** Naming Game. **(a)** Total number of words present in the system,  $N_w(t)$ ; **(b)** Number of different words,  $N_d(t)$ ; **(c)** Success rate  $S(t)$ , i.e., probability of observing a successful interaction at time  $t$ . The *inset* shows the linear behavior of  $S(t)$  at small times. The system reaches the final absorbing state, described by  $N_w(t) = N$ ,  $N_d(t) = 1$  and  $S(t) = 1$ , in which a global agreement has been reached. From Baronchelli et al. (2006b)

# AB Model

(minimal) naming game with only two possible words  $A$  &  $B$

rate of change can be calculated:

$$\dot{n}_A = -n_A n_B + n_{AB}^2 + n_A n_{AB}$$

$$\dot{n}_B = -n_A n_B + n_{AB}^2 + n_B n_{AB}$$

$$\dot{n}_{AB} = +2n_A n_B - 2n_{AB}^2 - (n_A + n_B)n_{AB}$$

fixed point solutions:

$$1 \quad n_A = 1$$

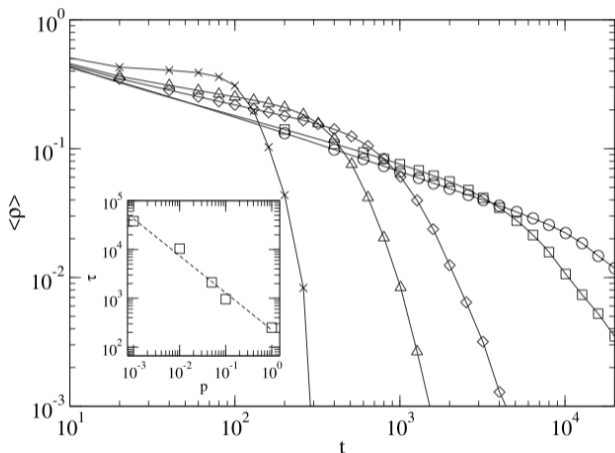
$$2 \quad n_B = 1$$

$$3 \quad n_A = n_B = 2n_{AB}$$

$n_X$  is the proportion of agents with vocabulary  $X$

## AB Model on SW-Networks

**Fig. 3** AB-model. Time evolution of the average density  $\langle \rho \rangle$  of bilingual individuals in small-world networks for different values of the rewiring parameter  $p$ . From *left to right*:  $p = 1.0, 0.1, 0.05, 0.01, 0.0$ . The *inset* shows the dependence of the characteristic lifetime  $\tau$  on the rewiring parameter  $p$ . The *dashed line* corresponds to a power law fit  $\tau \sim p^{-0.76}$ . From Castelló et al. (2006)



# Category Game

co-evolution of perceptual & linguistic categories

(for a continuous 1-dim perceptual space  $[0, 1]$ )

- population of  $n$  agents
- each agent  $i$  has:
  - a set of **categories**  $C_i$
  - for each  $c_j \in C_i$  a **vocabulary**  $V_{ij}$
  - for some  $c_j \in C_i$  a **designated word**  $d_{ij}$ 
    - last successful word, if exists
    - else the last one introduced, if exists
    - else none
- initially:
  - all  $C_i = \{[0; 1]\}$
  - all  $V_{ij} = \emptyset$
- asynchronous update with actual play (heterogeneous population)

(think: partition of  $[0, 1]$ )  
(set of words for  $c_j$ )

# Category Game: Play & Update Rule

## Preliminaries

- $C_i \subseteq [0; 1]$
- $V_i : C_i \rightarrow \mathcal{P}(\mathbb{N})$
- $C_i(a) = \min(\{z \in C_i \mid z > a\})$

(represent intervals by upper-bound)

(integers as words)

(category of  $a \in [0; 1]$ )

## Category Game: Play (1)

$i, j \leftarrow$  random speaker and hearer

$a, b \leftarrow$  random pair of perceptions from  $[0; 1)$  s.t.  $|a - b| > d_{\min}$

$a$  is the “topic” the speaker wants to talk about

# sender distinguishes stimuli if necessary

if  $C_i(a) = C_i(b)$ :

add  $a+b/2$  to  $C_i$

add  $\langle a+b/2, V_i(C_i(\max(a, b))) \rangle$  to  $V_i$

$w_1, w_2 \leftarrow$  random new words

add  $w_1$  to  $V_i(a+b/2)$

add  $w_2$  to  $V_i(C_i(\max(a, b)))$

$D_i(C_i(\frac{a+b}{2})) \leftarrow w_1$

$D_i(C_i(\max(a, b))) \leftarrow w_2$

( $i$ 's categories don't distinguish  $a$  and  $b$ )

(introduce new category boundary)

(new category inherits old vocabulary)

(add new random words)

(new words are distinguished)

## Category Game: Play (2)

# speaker chooses word  $w^*$  to send

if  $D_i(C_i(a))$  defined:

$$w^* \leftarrow D_i(C_i(a))$$

(if  $i$  has a distinguished word)

(choose distinguished word)

else:

$$w^* \leftarrow \text{uniform random from } V_i(C_i(a))$$

(choose random word)

# hearer collects possible interpretations

$$I \leftarrow \{x \in \{a, b\} \mid w^* \in V_j(C_j(x))\}$$

# hearer guesses intended referent

if  $I \neq \emptyset$ :

$$o^* \leftarrow \text{uniform random from } I$$

# determine success or failure

if  $I = \emptyset$  or  $o^* \neq a$ : failure

else: success

## Category Game: Update Rule

# hearer distinguishes objects if necessary

if  $C_j(a) = C_j(b)$ :

add  $a+b/2$  to  $C_j$

add  $\langle a+b/2, V_j(C_j(\max(a,b))) \rangle$  to  $V_j$

( $j$ 's categories don't distinguish  $a$  and  $b$ )

(introduce new category boundary)

(new category inherits old vocabulary)

# updating agents' vocabularies

if success:

(keep only  $w^*$ )

$V_i(C_i(a)) \leftarrow \{w^*\}$

$D_i(C_i(a)) \leftarrow w^*$

$V_j(C_j(a)) \leftarrow \{w^*\}$

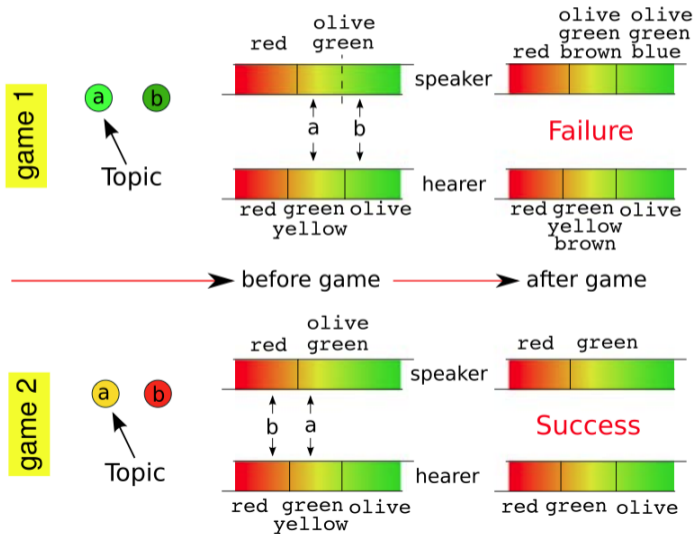
$D_j(C_j(a)) \leftarrow w^*$

else:

add  $w^*$  to  $V_j(C_j(a))$



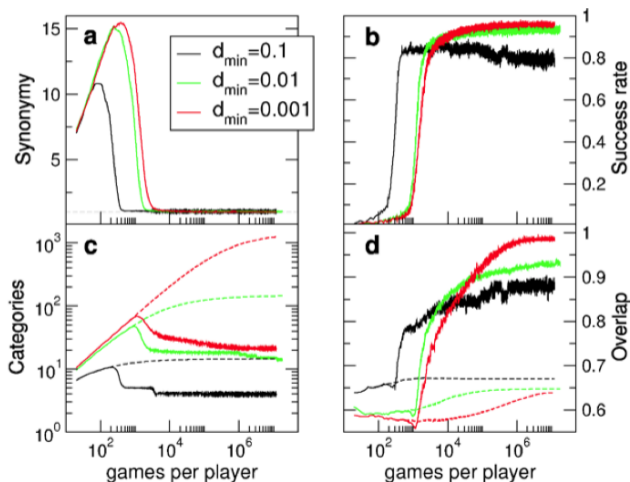
# Category Game: Example



(Loreto et al., 2010)

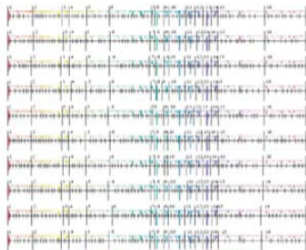
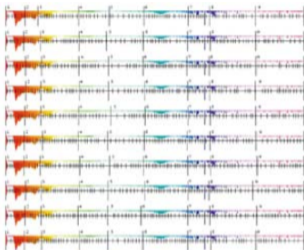
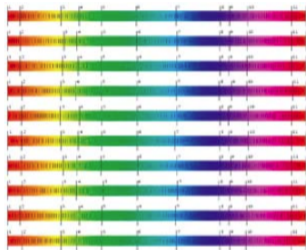
## Category Game: Results

**Fig. 5 Simulations results** with  $N = 100$  and different values of  $d_{\min}$ : **(a)** Synonymy, i.e. average number of words per category; **(b)** Success rate measured as the fraction of successful games in a sliding time windows games long; **(c)** Average number of perceptual (*dashed lines*) and linguistic (*solid lines*) categories per individual; **(d)** Averaged overlap, i.e., alignment among players, for perceptual (*dashed curves*) and linguistic (*solid curves*) categories



(Loreto et al., 2010)

# Category Game: Results



(Loreto et al., 2010)

## Numerical World Color Survey

compare two worlds: one with a uniform & one with a variable  $d_{\min}$

variable  $d_{\min}$  implements human JND for hue

uniform  $d_{\min}$  is set to .0143, the average of human JND

run 50 populations (50 agents each) in each world & look at resulting languages

compare simulation data against (subset of) data from **world color survey**

110 languages (without writing systems; small-scale, non-industrialized societies)

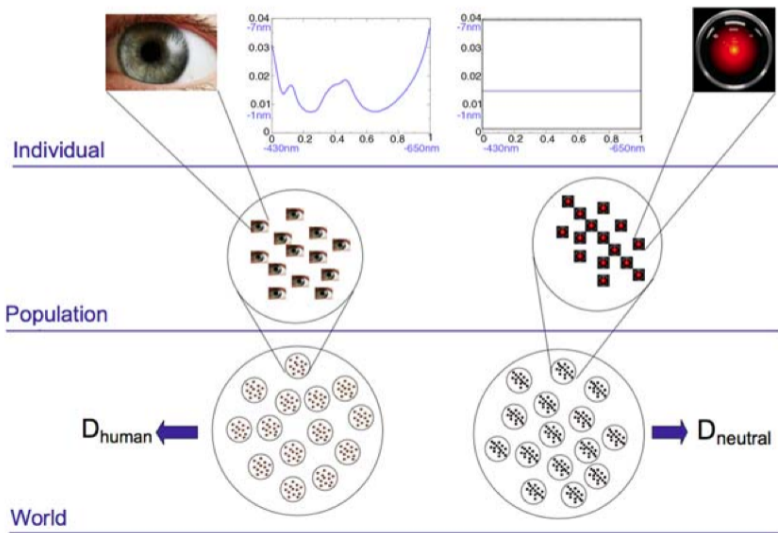
basic color term for each of 330 color chips for each language

ca. 24 speakers per language

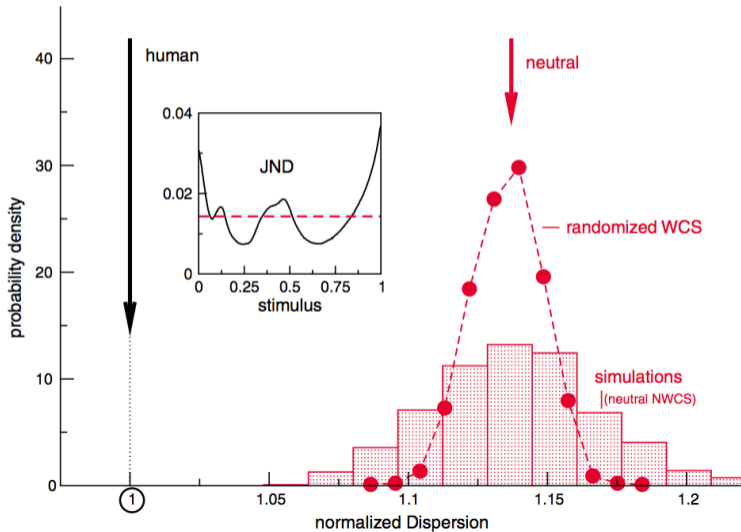
dispersion as a measure of common clustering (Kay and Regier, 2003):

$$D = \sum_{L, L'} \sum_{c \in L} \min_{c^* \in L'} \text{distance}(c, c^*)$$

# NWCS: Set-Up



# NWCS: Results



# Reading for Next Class

Michael Franke & Elliott Wagner (2014). "Game Theory and the Evolution of Meaning"  
*Language and Linguistics Compass* 8/9, 359–372

# References

- Baronchelli, Andrea et al. (2010). “Modeling the Emergence of Universality in Color Naming Patterns”. In: *PNAS* 107.6, pp. 2403–2407.
- Grim, Patrick et al. (2004). “Making Meaning Happen”. In: *Journal for Experimental and Theoretical Artificial Intelligence* 16, pp. 209–244.
- Kay, Paul and Terry Regier (2003). “Resolving the question of color naming universals”. In: *PNAS* 100.15, pp. 9085–9089.
- Loreto, Vittorio et al. (2010). “Mathematical Modeling of Language Games”. In: *Evolution of Communication and Language in Embodied Agents*. Ed. by Stefano Nolfi and Marco Mirolli. Springer-Verlag. Chap. 15, pp. 263–281.