Notes on Schelling's Model

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April 17, 2015

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1 Mark's Advice

- 1. Try to answer some philosophical question:
 - (a) To what extent can a simple model explain the real world?
 - i. What is life
 - ii. Is life complex?
 - iii. phase transitions?
 - iv. Is the behavior of the model exhibited in the real world?
 - (b) What kinds of emergence are possible?

2 Is the world chaotic

2.1 My Paper

- 1. Question: Is life Chaotic?
- 2. Proposed methodology: Explore a simple system that is a simplification of life.
 - (a) I will look at Schelling's Model and see what chaotic behavior exist in Schelling's model. I will look at the various metrics of chaos and see how they are in Schelling's Model.
 - i. Sensitivity to initial conditions
 - ii. Boundedness ecosystems/our world is bounded (yes and no)
 - A. Normal model: bounds on map so not really tested.
 - B. Try running Schelling's model in a big map while starting in a small area and see if agents reach the edge of the map.
 - iii. Lyapunov's exponent
 - iv. Neural Net prediction

v.

- 3. Important points about Schelling's Model:
 - (a) Parameters of Schelling's Model
 - i. Assume two races; lattice model; 4 neighbor vision; swapping? selection of who gets to move.
 - ii. Preferences: lower bound and upper bound.
 - iii. Size of map.
 - (b) Wolfram class?
 - (c) Turing Complete?

- (d) Behavior exhibited in the real world?
- (e) Local behavior?
- (f) Weak Emergence?
- (g) Self-Organization?

2.2My thoughts

- 1. Address the question: Is the world chaotic?
- 2. Chaos has various definitions:
 - (a) Sensitivity to initial conditions
 - (b) Stochastic in nature
 - (c) Random result
 - (d) Weak emergence
- 3. I want to explore the notion of weak chaos.
- 4. The intuition behind weak chaos is that...

2.3 Wolfram Paper

- 1. Direct mathematical analysis of 2D-CA is of little help
- 2. Resort direct simulation and random sampling.
- 3. Class 1,2,3,4
- 4. http://www.timswast.com/blog/2013/03/31/an-exploration-of-cellularautomata-as-dynamical-systems/
- 5. http://bactra.org/notebooks/cellular-automata.html
- 6. https://class.coursera.org/modelthinking/lecture
- 7. http://www.math.tamu.edu/mpilant/math614/
- 8. Notes from lecture:
 - (a) Some definition of dimension:
 - (b) Define orbit $\{x_i\}_{i=1}^N$.
 - (c) not that if $i \neq j$ then $x_i \neq x_j$ because consdiering noperiodic orbital
 - (d) In an orbit: define $val1 = \#\{(x_i, x_j) \mid d(x_i, x_j) < \epsilon, i \neq j\}$.
 - (e) $val2 = \#\{(x_i, x_j) \mid i \neq j\} = \frac{N(N+1)}{2}$ (f) ratio $= c(\epsilon) = \frac{val1}{val2}$

 - (g) $\epsilon \to \infty \implies c(\epsilon) \to 1$.
 - (h) $\epsilon \to 0 \implies c(\epsilon) \to 0$.
 - (i) correlation exponent = $\alpha = \lim_{\epsilon \to 0} \left| \frac{\ln c(\epsilon)}{\ln \epsilon} \right|$

- (j) Correlation exponent defined in terms of orbit; not the space.
- (k) Therefore the correlation exponent is easy to calculate.
- (1) Loop over N(N+1)/2 pairs of x_i, x_j where $i \neq j$.
- (m) calculate $d(x_i, x_j)$. If $dist < \epsilon$, increment counter
- (n) implies n^2 operations. implies very efficient.
- (o) $N(\epsilon)$ =number of boxes of width ϵ that contain a point in orbit.
- (p) define α the same way for $N(\epsilon)$.
- (q) Haussdorff Dimension
- (r) given set S. cover with balls of diamter ϵ .
- (s) define $H_m(S) = \lim_{\epsilon \to 0} \sum_{\text{number balls}} (\epsilon)^m$
- (t) example curve of length L takes L/ϵ balls of diamter ϵ so $H_1(S) = \lim_{\epsilon \to \infty} (L/\epsilon)\epsilon = L$.
- (u) $H_2(S) = \lim_{\epsilon \to 0} (L/\epsilon) \epsilon^2 = 0$.
- (v) So Haussdorf dimension = largest d such that $H_d(S) \neq 0$.
- (w) equivalent to smallest d such that $H_d(S) = 0$.
- (x) expensive to compute.
- (a) Question: when is an orbit chaotic?
- (b) Take finite orbit $\{x_i\}_i^N$. and orbit z_i
- (c) $d(x(t), z(t)) \sim d_o e^{\alpha t} \Longrightarrow \text{chaotic if } \alpha > 0.$
- (d) Lyapunov exponent = $\alpha = \frac{1}{1} \ln |\frac{x(t) z(t)}{x_0 z_0}|$
- (e) Different orbits can have different exponents.
- (f) Also might be dependent on component/dimension that you are looking (x-component verse y-component)
- (g) Wolf Algorithm (1983) for computing approximate lyapunov exponents of an orbit:
 - i. Pick initial point x_0 .
 - ii. Iterate till close to attractor. Say choose N >> 1
 - iii. Fix x_N , find another point (x_m) in the orbit such that x_m is close to x_N . i.e. $d(x_m, x_N) \sim 10^4 \epsilon$ where ϵ is machine ϵ i.e. smalleest floating point number.
 - iv. i.e. $\epsilon = 10^{-16}$ in double precision. 10^{-8} in single precision.
 - v. do stuff with sequences. slide 10/11

3 Thesis

Introduction

1. Check out sugarscape model.j

- (a) An agent based model for Schelling
- (b) from 1969. There exists java implementations somewhere.
- 2. Bak et al. use statistical mechanics upon he self-organized systems.
- 3. Bak: global emergent properties have the property that they are scalefree.
- 4. Author follows that with the claim: Therefore we can use global emergent properties to measure the system.
- 5. Zhang uses Game Theory to model Schelling' Model
- 6. each agent has fixed income. each agent has π that correlates to the happiness of an agent.
- 7. Gives locations a price, number white nbrs, number black nbrs. Then uses simply supply and demand to determine prices
- 8. page 21 discusses some research on Schelling's model
- 9. Schelling focuses on preferences of agents
- 10. other research on the neighborhood of agents, up to a distance R, calling this parameter an agent's vision
- 11. Herb Simon's 'situated ant'
- 12. An ant following a random walk on a beach produces a complex path. But it is the complexity of the environment that is creating the complex path rather than the ant
- 13. So complexity of environment is as important as complexity of individual.
- 14. My Analysis of section:
 - (a) Author points out a variety of systems:
 - i. Agent Based Model (sugarscape)
 - ii. Statistical Mechanics (sandpiles)
 - iii. Game Theory (Zhang's Model with happiness and incomes)
 - iv. Complex Environment verse Complex Ant

3.1 On Schelling

- 1. Three models:
- 2. Spatial Proximity Model
- 3. Bounded Neighborhood Model = characters consider proportions within system rather than in a local neighborhood of the system.
- 4. Tipping Model
- 5. Author says Schelling's Models too simplistic to model real world
- 6. Author says need to make environment more complex

3.2 Further Work

- 1. Heterogenous Populations: explore different types within population. (e.g. each white person has a different race tolerance)
- 2. Space: Explore nonuniform spaces. Change perceptions of Space
 - (a) This is done to some extent in the thesis.
 - (b) The author explores a model where agents have a random variable for their "vision", i.e. they can see all neighbors with r spaces from them where r is a random variable.
- 3. Timescales: Explore the effects of agents moving at different timescales, either individually or by type.
- 4. Time decay (or growth) for social bonds: would a social bond that decayed (or strengthened) over time affect the results of the model?
- 5. Ratio Calculations: Instead of agents recalculating ratio every iteration, only calculate it every once in a while.
- 6. Population Adaptation: have the population change over time. This turns Schelling's model into an evolutionary model
- 7. Little work on Bounded Neighborhood Model
- 8. Stoica and Flache apply Schelling's residential segregation to school segregation by incorporating Zhang's utility (measuring distance from a school instead of price)