

Final Project Progress Report # 3

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Introduction

People claiming no religious affiliation constitute the fastest growing religious minority in many countries, including the United States [1]. In fact, the religious “nones” are the only group growing in all 50 US states [2]. Although many scholars attribute the decline of religious affiliation to generational changes, roughly half of the US population changes religious affiliation at some point in their life, often several times [3]. This suggests that religious affiliation shift can be modeled as social group competition, with different religious groups competing for members [4]. Such models predict that religious coexistence is not a stable state; the unaffiliated group will grow until all organized religion has disappeared. Whether or not this comes to fruition, all people have a stake in the outcome.

Dynamical systems model

Abrams et al. split an ideal society into the mutually exclusive religiously affiliated and unaffiliated, with the fraction x belonging to the unaffiliated group and $y = 1 - x$ belonging to affiliated. Assuming that people only switch affiliations based on the fraction of people in each group and the perceived utility of the group, the dynamics of conversion can be modeled by

$$\frac{dx}{dt} = yP_{yx}(x, u_x) - xP_{xy}(x, u_y), \quad (1)$$

where $P_{yx}(x, u_x)$ is the probability per unit time that an individual converts from religious to unaffiliated, $0 \leq u_x \leq 1$ is the perceived utility of being unaffiliated, and $u_y = 1 - u_x$ is the perceived utility of being affiliated. The authors further assume that (1) is symmetric under exchange of x and y and that no individual would switch to a group with no members (i.e. $P_{yx}(0, u_x) = 0$).

If the transition probabilities are smooth and monotonically increasing in both arguments, then there exist at most three fixed points with alternating stability. All available data suggest that the inevitable steady state is $x^* = 1$, or the extinction of religious affiliation. For specificity, the authors chose the power law $P_{yx}(x, u_x) = x^a u_x$; the best fit to data occurs for $a = 1$. The authors extend this model to binary networks of individuals (rather than all-to-all coupling) and allow for a continuous “religiosity” degree (rather than binary in or out of group), but the final state remains the same. As long as the network is not completely disconnected, only a time delay is introduced.

Agent based model

Minimal continuous dynamical systems lend themselves well to rigorous analysis, but important details may be left out for simplicity. Assuming religious affiliation change can be modeled as social group competition, agent based modeling is a natural way to test the robustness of the continuous model results.

Previous versions

The previous versions verified that an agent based implementation of (1) with an well-connected network replicates the approximately logarithmic growth of the unaffiliated group and the eventual extinction of the religiously affiliated. If NETWORK? is switched on, people change affiliations based only on their friends' affiliations rather than the affiliated proportion of the entire population. The social network also changes as people switch affiliations, detailed below.

Initially, COMMUNITY-SIZE agents are given a random unaffiliated utility sampled from a normal distribution with mean U-X and standard deviation U-X-VAR (taken to be 0 for initial tests). A proportion INITIAL-X of the population is unaffiliated and the rest are affiliated. To be consistent with all available data, INITIAL-X must be small ($\approx 5\%$). If NETWORK? is on, a spatially clustered network with AVERAGE-NODE-DEGREE is created.

At each tick, affiliated agents switch to unaffiliated with probability $\text{TIME-SCALE} * u_x * x^{\text{AFFILIATION-POWER}}$, where x is the unaffiliated proportion of an agent's friends. Unaffiliated agents switch to affiliated with probability $\text{TIME-SCALE} * (1 - u_x) * (1 - x)^{\text{AFFILIATION-POWER}}$. See Figure 1.

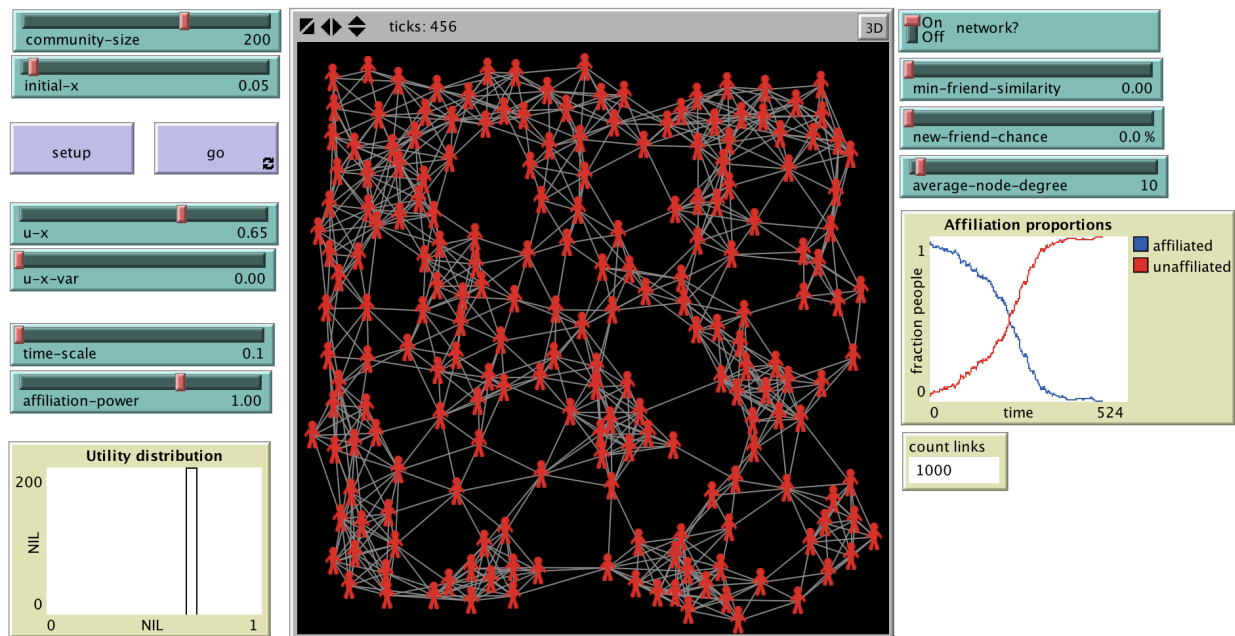


Figure 1: Verification that the agent based implementation of (1) on a network replicates the logarithmic growth of an initially small unaffiliated population. For a uniform unaffiliated utility u_x exceeding 0.5, the religiously affiliated group will convert to unaffiliated, but with a time delay from the all-to-all network.

As seen in [4], the unaffiliated group grows approximately logarithmically, and the affiliated group eventually disappears. When the network is not all-to-all, the system reaches steady state with a time delay, which is also consistent with the dynamical systems model. As AVERAGE-NODE-DEGREE decreases, the time delay increases. For small average node degree (< 5), the affiliated group no longer goes extinct because small clusters are not connected.

After confirming that this version of the agent based model produces the same results as the dynamical systems model on a network, I continue to test the robustness of the model results.

Add affiliation-dependent network

One of the assumptions of the continuous model (1) is that the network is static, but it's more realistic to assume that people lose and make friends when they switch affiliations. At each time step, people switch affiliations according to the previous transition probabilities. However, if an agent switches from affiliated to unaffiliated (or vice versa) in that time step, it will lose some affiliated friends, but it may gain a new unaffiliated friend.

Each time an agent switches from affiliated to unaffiliated, it calculates what proportion of its friends are unaffiliated. If that proportion is less than the MIN-FRIEND-SIMILARITY desired by all agents, the agent break ties with random affiliated friends until the MIN-FRIEND-SIMILARITY proportion is reached. If agents want most of their friends to have similar affiliations, then the unaffiliated and affiliated groups can coexist. The system stabilizes when the affiliated agents either form their own separate cluster or disconnect from the network entirely. See Figure 2.

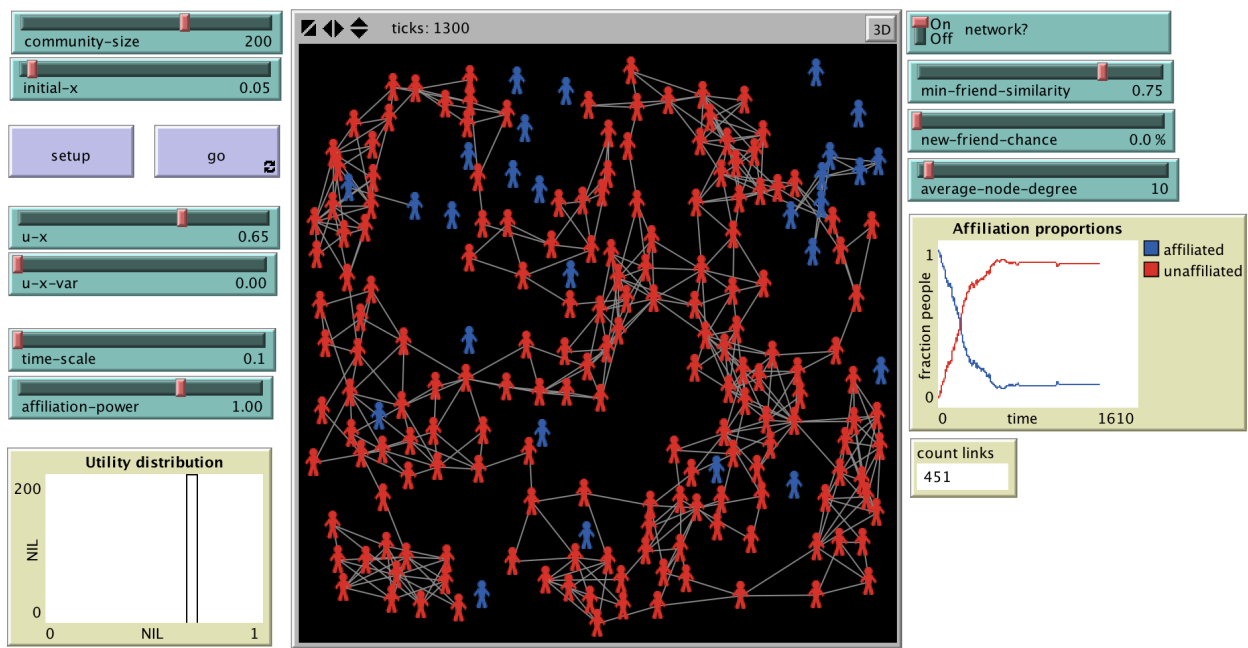


Figure 2: When the agents want to be friends with at least 75% similarly affiliated agents, the two groups can coexist. At equilibrium, affiliated agents have either formed their own separate cluster or disconnected from the network entirely.

In addition to losing friends after switching from affiliated to unaffiliated, each agent might gain a new unaffiliated friend with NEW-FRIEND-CHANCE. As this probability increases, the proportion of unaffiliated agents at equilibrium increases. See Figure 3.

Combine all previous version features

I have thus far tweaked model parameters one at a time. If for instance I add variation to affiliation utility, shift the affiliation power off of exact $a = 1$, set the desired friend similarity to 66%, and the chance of making new friends to 10%, the agent based model results are consistent with real-world data but the equilibrium state disagrees with the continuous model predictions. See Figure 4.

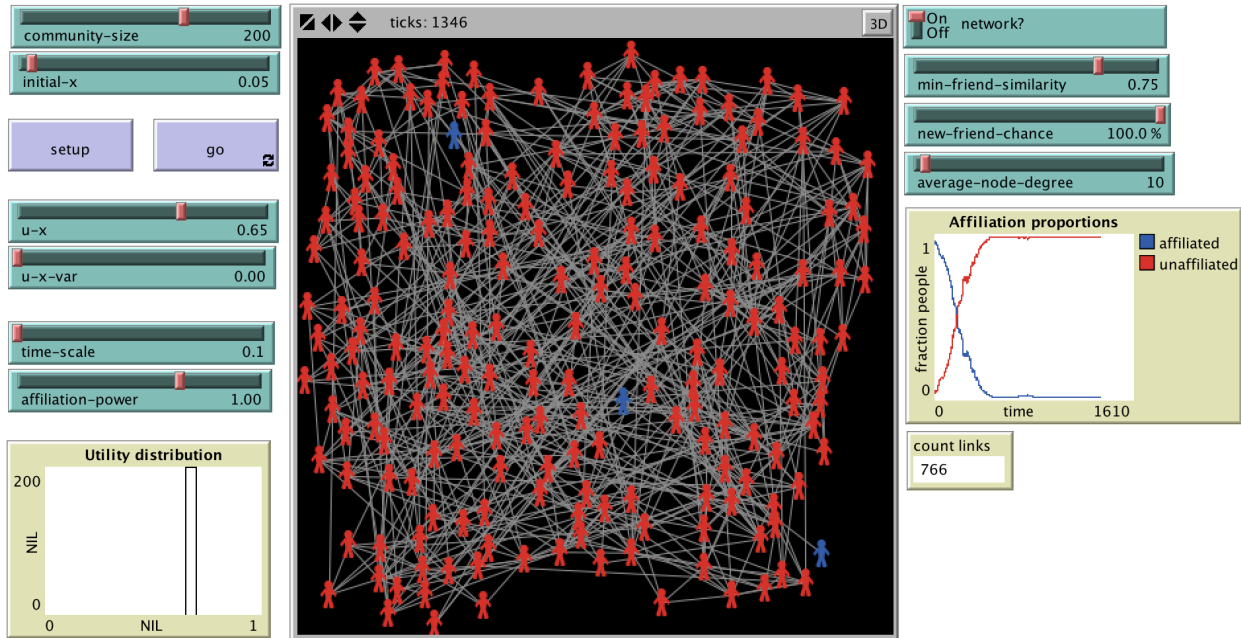


Figure 3: When the agents have a good chance of making friends with similarly affiliated agents after switching groups, the two groups barely coexist. Only a few affiliated agents remain, often unattached to anyone in the network.

Current version

The current version resolves several known bugs from the previous version and reveals more potential bugs. This week, I cleaned and commented the code, conforming with the style guide. I broke large chunks of code into smaller modules, allowing me to verify that the model behaves as intended. It also allowed me to reconsider how the agents make friends. Finally, I rearranged the interface to conform to the style guide and displayed the network in a more meaningful way.

As in previous versions, `COMMUNITY-SIZE` agents are initially given a random unaffiliated utility sampled from a normal distribution with mean `U-X` and standard deviation `U-X-VAR` (taken to be 0 for initial tests). A proportion `INITIAL-X` of the population is unaffiliated and the rest are affiliated. If `NETWORK?` is on, a spatially clustered network with `AVERAGE-NODE-DEGREE` is created.

At each tick, affiliated agents switch to unaffiliated with probability $\text{TIME-SCALE} * u_x * x^{\text{AFFILIATION-POWER}}$, where x is the unaffiliated proportion of an agent's friends. Unaffiliated agents switch to affiliated with probability $\text{TIME-SCALE} * (1 - u_x) * (1 - x)^{\text{AFFILIATION-POWER}}$.

Each time an agent switches from affiliated to unaffiliated, it calculates what proportion of its friends are unaffiliated. If that proportion is less than the `MIN-FRIEND-SIMILARITY` desired by all agents, the agent break ties with random affiliated friends until the `MIN-FRIEND-SIMILARITY` proportion is reached.

In addition to potentially losing friends after switching from affiliated to unaffiliated, each agent might gain a new friend proportional to his loneliness at each tick; in other words, the agent will gain a friend with probability $\text{NEW-FRIEND-CHANCE} / (\text{number-friends} + 1)$. If an agent already has `MAX-FRIENDS`, it won't attempt to make more friends. It seems more realistic that people make connections somewhat randomly, but break those connections based on different affiliations. See Figure 5.

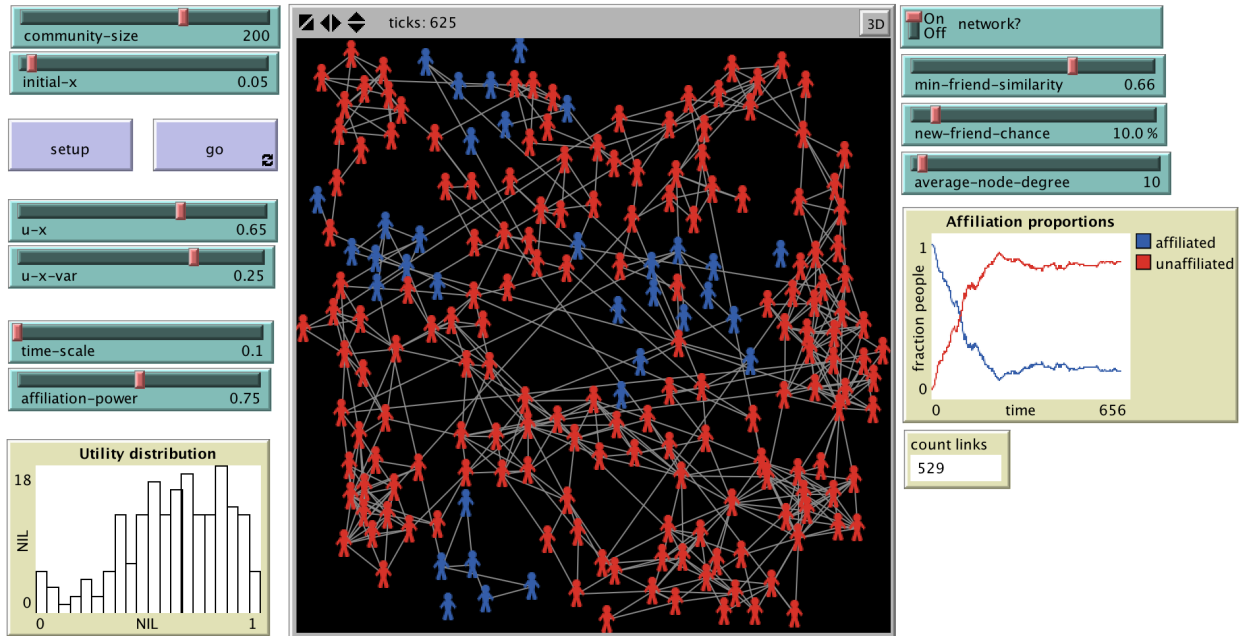


Figure 4: Changing all model parameters at once, I see a consistent pattern. Consistent with real-world data, the number of unaffiliated agents grows until the the unaffiliated group is the clear majority. Then the number of unaffiliated members peaks before settling down into a slightly lower proportion. At equilibrium, the two groups coexist with very few alienated individuals. This seems like a realistic outcome consistent with real-world data.

Model verification

When I broke the code into smaller modules, I fixed the known bug from last week (algorithm does not perfectly meet agents' desire to have a least MIN-FRIEND-SIMILARITY). I also discovered and fixed a bug that allowed agents to stay with their current affiliation even when they intended to switch.

I found but have not addressed a potential bug with breaking friendships. Currently, agents look at the proportion of their friends who are affiliated at each tick in order to decide how many friends to break ties with. It may be a problem that the agentset of friends is constant during a tick (unless a link is broken), but the affiliations of those friends may change during the tick. This is inconsistent with my intentions, but I don't know if it's a problem.

Next steps

The following additions and verifications will be conducted next:

1. The execution of breaking friendships is inconsistent with my intentions. I will modify the algorithm to conform to my intentions, though I doubt it will behave qualitatively differently.
2. I will validate the model with the available data. With certain "reasonable" parameters, the model must fit the data for each country. The data cannot validate the equilibrium state of the model because the real world is not close to equilibrium. All countries have less than 30% unaffiliated populations.

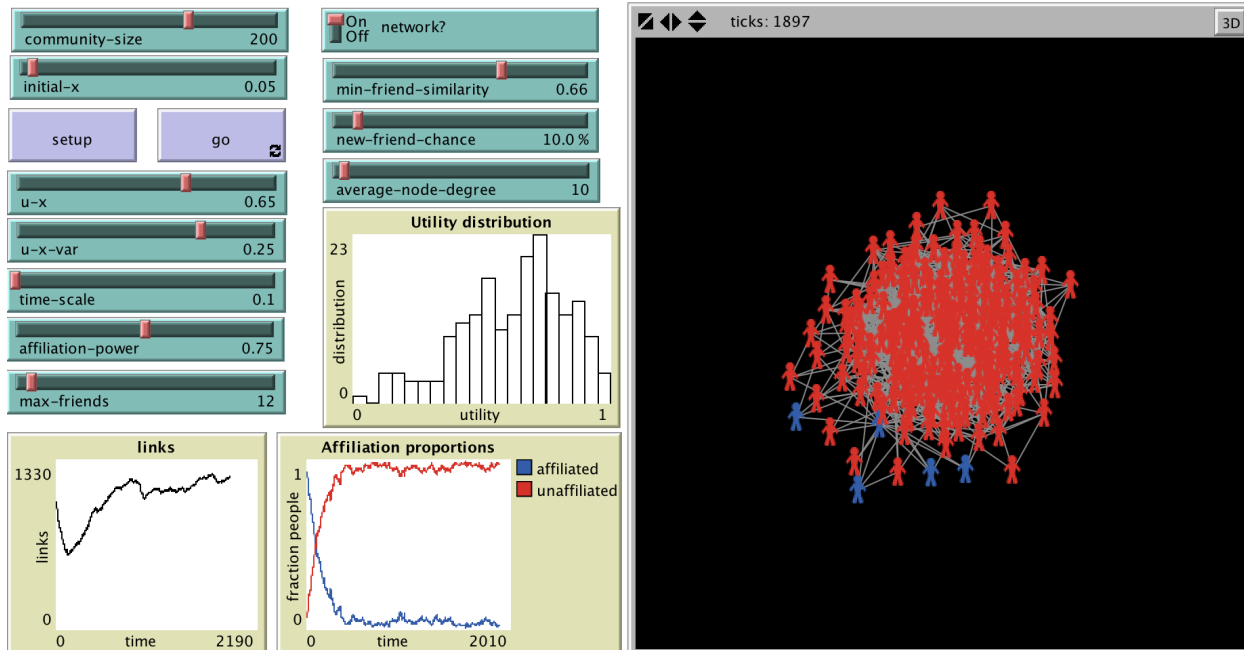


Figure 5: When the agents make friends proportional to their loneliness (rather than seeking friends with similar affiliations), agents don't disconnect from the network and become hermits forever. This causes the unaffiliated group to grow larger than before (Fig. 4) using the same settings. However, the affiliated group does not go extinct.

3. I will build a generic social group competition model using HubNet where users choose a group to join based on the current membership and utility of the group. It would be interesting to see if the same transition functions emerge from this abstraction of group competition. I imagine that large patches would represent affiliations, and users would move to those patches based on provided utilities (varying across users) and the number of other agents on the same patch. If I want to impose a network structure, non-neighboring agents would be invisible.
4. Time permitting, I will incorporate birth, death, and immigration into the model. Different affiliations may have different birth and death rates. Immigrants may have different religious affiliations from the current population.

Conclusions

The third version of the agent based model of religious affiliation replicates the continuous dynamical systems model on a network under the specified conditions. Specifically, the agent based model predicts logarithmic growth of the unaffiliated group and the eventual extinction of the affiliated group. The early dynamics of the model are consistent with real-world data, but the world is far from religious equilibrium. It is therefore impossible to use real-world data to select which of the models best describes religious shift or what the eventual outcome will be.

I am testing the robustness of the continuous model predictions under various reasonable conditions. After allowing people to perceive different utilities for religious affiliation (while still retaining the average utility implied by the data), coexistence of the affiliated and unaffiliated is possible. Tweaking the exponent

of the power law switching probability also allowed coexistence. These realistic additions to the model are still consistent with real-world data, but the final state can be qualitatively different.

Coexistence of the affiliated and unaffiliated groups is also possible when the people are connected in a social network, only looking to their friends to decide their affiliation. When people break ties with differently affiliated people and make friends somewhat randomly, the unaffiliated group becomes a clear majority, but a small number of affiliated members survive. Therefore the extinction of religious affiliation is not inevitable.

References

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- [4] D. M. Abrams, H. A. Yapple, and R. J. Wiener, "Dynamics of social group competition: modeling the decline of religious affiliation," *Physical Review Letters*, vol. 107, no. 8, p. 088701, 2011.