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ABSTRACT

Tipping and Residential Segregation: A Unified Schelling Model*

This paper presents a Schelling-type checkerboard model of residential segregation formulated as a spatial game. It shows that although every agent prefers to live in a mixed-race neighborhood, complete segregation is observed almost all of the time. A concept of tipping is rigorously defined, which is crucial for understanding the dynamics of segregation. Complete segregation emerges and persists in the checkerboard model precisely because tipping is less likely to occur to such residential patterns. Agent-based simulations are used to illustrate how an integrated residential area is tipped into complete segregation and why this process is irreversible. This model incorporates insights from Schelling’s two classical models of segregation (the checkerboard model and the neighborhood tipping model) and puts them on a rigorous footing. It helps us better understand the persistence of residential segregation in urban America.

JEL Classification: C72, C73, D62, R13

Keywords: residential segregation, tipping, checkerboard model

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1. Introduction

For three decades, survey data have consistently shown that African Americans prefer to live in integrated neighborhoods with half blacks or a slight black majority.\(^1\) For example, in 1982, the General Social Surveys asked black respondents about their most preferred neighborhood racial composition and found that 61.6 percent of them picked “half black, half white” as the top choice (Davis and Smith, 1993). Data collected in a Multi-City Study of Urban Inequality (MCSUI) during the 1990s also showed that 50 percent of black interviewees chose a 50-50 neighborhood as the most attractive and 99 percent of them indicated a willingness to move into such neighborhoods (Krysian and Farley, 2002).

Survey results also demonstrated that whites, although generally less enthusiastic about 50-50 type neighborhoods than blacks, have increasingly endorsed the idea of residential integration (Schuman et al., 1997). The same MCSUI data indicated that 60 percent of whites felt comfortable with neighborhoods with one-third blacks and that 45 percent of whites were willing to move into such neighborhoods (Charles, 2003).

These stated preferences are in stark contrast with the high levels of racial housing segregation in reality. Using the 1990 Census data, Massey and Denton (1993) find that residential segregation between blacks and whites remains a striking feature of many U.S. metropolitan areas, a situation they describe as the “American apartheid.” According to the same Census data, very few blacks actually live in “half black, half white” neighborhoods, in contrast to the survey results cited above (Zhang, 2004a).\(^2\) The latest 2000 Census data show only modest steps toward integration (see, e.g., Glaeser and Vigdor, 2001; Logan et al., 2004).\(^3\)

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\(^1\) See, e.g., Farley et al. (1978), Farley et al. (1994), Farley et al. (1997), Krysan and Farley (2002), Schuman et al. (1997), and Bobo and Zubrinsky (1996).

\(^2\) For example, only 2.91 percent of blacks in Chicago, 2.97 percent in Cleveland, and 2.56 percent in Detroit live in “half black, half white” neighborhoods, which Zhang (2004a) broadly defines as any neighborhood with 40-60 percent blacks.

\(^3\) In terms of long-term trend, the absolute degree of residential segregation peaked in 1970 and has been declining since (Cutler et al., 1999; Glaeser and Vigdor, 2001). However, the decline is too small relative to the dramatic change in racial attitudes, federal housing policy, and the socioeconomic status of blacks.
This persistence of residential segregation in urban America is generally considered a social problem because it has adverse effects, especially on blacks. Kain (1968) has long argued that residential segregation leads to a “spatial mismatch” between blacks and their potential employers that diminishes blacks' employment opportunities. Wilson (1987) echoes this point in a highly influential study of the “truly disadvantaged” inner city underclass in America. Massey and Denton (1993) argue that segregation is at the root of many problems facing blacks. Numerous studies have since documented that residential segregation is associated with negative outcomes for blacks in terms of academic performance, educational attainment, employment, single parenthood, poverty, criminal behaviors, and even health outcomes (see, e.g., Card and Rothstein, 2007; Cutler and Glaeser, 1997; Galster, 1987; Orfield and Eaton, 1996; Shihadeh and Flynn, 1996; Williams and Collins, 2001).4

There are three primary explanations for the persistence of segregation.5 The first is derived from a long-standing hypothesis that physical distance between racial groups reflects their social distance. This explanation holds that segregation persists because lower-income blacks cannot afford to move to white neighborhoods although they would like to (Clark and Ware, 1997; Cloutier, 1982; Iceland and Wilkes, 2006; Marshall and Jiobu, 1975).6 The second explanation emphasizes various forms of racial discrimination in the housing market. It argues

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4 To what extent these negative effects are causal is still debatable. For example, evidence from a recent randomized policy experiment (HUD’s Moving to Opportunity program) shows that moving out of high-poverty neighborhoods has limited positive effects on minority families. See Kling et al. (2007) and many of the evaluation studies they cite. See also Ludwig et al. (2008) and Clampet-Lundquist and Massey (2008) for a debate on the validity of evidence from the experiment.

5 Dawkins (2004) points out a fourth possible explanation of the persistence of segregation that stresses on the information asymmetry in the housing market. More specifically, if information on housing mostly comes from local sources, then blacks living in predominantly black neighborhoods would have very limited information about the availability and costs of housing in predominantly white neighborhoods. This lack of information may prevent them from moving across the color line. This hypothesis, though equally plausible, has been under-researched.

6 This hypothesis has always been challenged because empirical evidence has consistently shown that high levels of segregation exist in every socioeconomic group (see, e.g., Taeuber, 1968; Farley, 1995; Darden and Kamel, 2000; and Zhang, 2003). Bayer et al. (2004) find that socioeconomic factors explain only a small proportion of the segregation of blacks although they can explain a large proportion of the segregation of Hispanics.
that discriminatory behaviors of real estate agents, mortgage lenders, and insurance providers prevent blacks from moving into their most preferred neighborhoods (see, e.g., Goering and Wienk, 1996; Yinger, 1986, 1995). The third explanation points to the fact that whites have less favorable feelings toward integrated neighborhoods than blacks. It contends that there are not enough whites willing to reside in neighborhoods with 50 percent or more blacks (Emerson et al., 2001; Krysian and Farley, 2002; Vigdor, 2003).\footnote{These three explanations are not necessarily mutually exclusive. For example, Sethi and Somanathan (2004) explain residential segregation based on a preference over both neighborhood affluence and its racial composition. Their model suggests that under certain conditions a narrowing racial income gap may increase rather than decrease the levels of segregation.} Given these explanations for the persistence of segregation, one naturally expects that residential integration will occur as long as socioeconomic inequalities between blacks and whites continue to narrow, whites’ attitudes toward integrated neighborhoods continue to become more favorable, and anti-discrimination laws are more stringently enforced.\footnote{In fact, there is evidence for the changing attitudes of whites. Survey data show that more and more whites tend to agree with the principle that blacks should be able to live wherever they can afford and that more and more whites tend to approve open-housing laws (Schuman et al., 1997; Bobo, 2001). Time-series data from the Detroit area also show that whites are more and more willing to remain in their neighborhoods as blacks enter (Farley et al., 1994). This change in attitudes alone could lead to a decline in racial discrimination. Ross and Turner (2005) find that indeed discriminatory behaviors have become less pervasive in U.S. housing markets.}

In this paper, I show that the prospect of residential integration may be bleaker than what is generally perceived. I present a rigorous model that incorporates important insights from Thomas Schelling’s two models of segregation (Schelling, 1969, 1971, 1972, and 1978). I will show that residential segregation could emerge and persist even if the following three conditions hold simultaneously: (1) There exists no racial discrimination of any type; (2) both blacks and whites prefer to live in 50-50 neighborhoods; and (3) the socioeconomic disparities between blacks and whites are completely eliminated.

This paper makes two contributions. First, it offers an explanation for the persistence of residential segregation in U.S. metropolitan areas observed in the past two decades. Existing studies on the persistence of segregation generally assume that if integration is a preferred
outcome, then segregation could persist only if some frictions prevent the transition to integration. This is why many researchers consider racial discrimination and income disparities as possible reasons for the persistence of segregation. Deviating from this literature, this paper emphasizes that racial integration, even if highly desirable at the individual level, may not be attainable because integrated residential patterns are inherently unstable. This idea is well-known because of Schelling’s pioneering studies and follow-up work, but it has not been a focus of recent research on the persistence of segregation.

Second, this paper formalizes Schelling’s idea of tipping and uses it to rigorously demonstrate why residential segregation may emerge and persist despite the preference for integration at the individual level. Schelling originally developed the idea of tipping to study segregation dynamics in a single neighborhood and the concept has always been discussed in the context of a single neighborhood. Schelling also showed, in a multi-neighborhood setting, that residential patterns may not reflect individual preferences. By introducing the idea of tipping into a multi-neighborhood model, this paper incorporates the insights from Schelling’s separate models into a unified framework. Unlike Schelling’s inductive and less formal approach, this paper presents a rigorous model and analyzes it using techniques from mathematical game theory, which helps sharpen many of the important insights developed by Schelling.

The remainder of the paper is organized as follows. Section 2 reviews Schelling’s two dynamic models of segregation. Section 3 presents a unified Schelling model that combines the insights from both of his models. Section 4 uses agent-based simulations to illustrate the analytical results of the unified model. Section 5 concludes with some remarks.

2. Schelling’s Models of Segregation

In a series of publications, Schelling (1969, 1971, 1972, and 1978) presents two distinct models of residential segregation between blacks and whites: the spatial proximity model and the
bounded-neighborhood model. Each model provides great insights into the causes and the persistence of racial housing segregation. Both are well known and highly influential among social scientists far beyond the research area of segregation.

The spatial proximity model is a simulation of the dynamics of segregation. It starts with a residential area (a line or a grid) populated by a fixed number of black and white agents, each having a preference over the racial composition of her immediate neighborhood. Agents dissatisfied with their current residential locations are allowed to move into their preferred neighborhoods in a random order. The simulation stops when no agent is found to be discontent. Schelling’s most striking finding is that moderate preferences for same-color neighbors at the individual level can be amplified into complete residential segregation at the macro level. For example, if every agent requires at least half of her neighbors to be of the same color—a preference far from extreme—the final outcome, after a series of moves, is almost always complete segregation. Thus Schelling concluded that the “macrobehavior” in a society may not reflect the “micromotives” of its individual members.

The spatial proximity model has every ingredient of a great theoretical work: it addresses an important real world issue; it is simple to understand; it produces unexpected results; and it has deep implications for various social sciences. In addition, the model also has the appealing feature that any reader can come up with a variation of its two-dimensional version and simulate it, for example, by moving dimes and pennies on a checkerboard. This model has since become widely known as Schelling’s “checkerboard model.” Because of this work, many researchers today consider Schelling to be a pioneer in agent-based computational economics (Epstein and Axtell, 1996, p. 3). The model’s numerous variations and the robustness of its main result have

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9 Schelling (1972) focuses on the bounded-neighborhood model only and calls it the “neighborhood tipping” model. Each of the other works covers both models.

10 Schelling was awarded the 2005 Nobel Prize in economics “for having enhanced our understanding of conflict and cooperation through game-theory analysis.” Although the Nobel citation focused on his work The Strategy of Conflict (Schelling, 1960), both his models of residential segregation were also recognized as his important contribution to the social sciences (The Royal Swedish Academy of Sciences, 2005a, 2005b).
been a popular topic in the area of agent-based modeling (see, e.g., Bruch and Mare, 2006; Epstein and Axtell, 1996; Fagiolo et al., 2007; Fossett, 2006; O'Sullivan, 2008; Pancs and Vriend, 2007).¹¹

Despite the important insight revealed in Schelling’s simulation of the checkerboard model, for many years social scientists were unable to rigorously analyze the model, primarily because of the lack of suitable mathematical tools. Young (1998) was the first to point out that techniques recently developed in evolutionary game theory—particularly the concept of stochastic stability introduced by Foster and Young (1990)—are useful for analyzing Schelling’s spatial proximity model. Young (1998, 2001) presents a simple variation of the one-dimensional Schelling model (on a ring). He shows that segregation tends to appear in the long run even though a segregated neighborhood is not preferred by any agent.¹² Pancs and Vriend (2007) derive similar results showing that complete segregation is the only possible long-run outcome on a ring when agents with a preference for 50-50 neighborhoods play best response to neighborhood racial composition.¹³ Formulating the checkerboard model as a spatial game played on a two-dimensional lattice graph, Zhang (2004a) shows that even if everybody prefers to live in a 50-50 mixed-race neighborhood, complete segregation emerges and persists in the long run under fairly general conditions. All of these results are even stronger than those Schelling showed in his original model. In another paper, Zhang (2004b) enriches the checkerboard model by adding a simple housing market. It not only offers an alternative account of residential segregation, but also produces testable hypotheses on housing price and vacancy differentials between predominantly black and predominantly white neighborhoods. These studies all seek to


¹² Bog (2006) assesses the robustness of the results from Young’s one-dimensional model.

¹³ This analytical result in Pancs and Vriend (2007) cannot be extended to a two-dimensional setting, but using simulations they show that best-response dynamics tend to produce segregation even in a two-dimensional space.
reformulate Schelling’s checkerboard model in a game-theoretic framework and place it on a rigorous footing. The goal is to apply standard results in game theory to gain a better understanding of the insights illustrated in Schelling’s checkerboard model.14

Schelling’s second model of residential segregation, the “bounded-neighborhood model,” is better known as the neighborhood tipping model.15 Unlike the checkerboard model, which is a simulation, the tipping model is purely analytical. It seeks to explain the phenomenon of neighborhood tipping. Tipping is said to have occurred when an all-white neighborhood, after some blacks move in, suddenly begins the process of evolving into an all-black neighborhood with more and more whites moving out and only blacks moving in. This process of tipping, once started, often appears to be accelerating and irreversible. Schelling’s explanation is based on the assumption that different residents in a neighborhood have different tolerance levels toward the presence of neighbors of the opposite color. For example, in an all-white neighborhood, some residents may be willing to tolerate a maximum of 5 percent black neighbors; others may tolerate 10 percent, 20 percent, and so on. The ones with the lowest tolerance level will move out if the proportion of black residents exceeds 5 percent. If only blacks move in to fill the vacancies after the whites move out, then the proportion of blacks in the neighborhood may reach a level high enough to trigger the move-out of the next group of whites who are only slightly more tolerant than the early movers. This process may continue and eventually result in an all-black neighborhood.

Similarly, an all-black neighborhood may be tipped into an all-white neighborhood, and a mixed-race neighborhood can be tipped into a highly segregated one, depending on the tolerance

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14 Empirical work motivated by Schelling’s checkerboard model seems rare. Ruoff and Schneider (2006) is an exception that tests the Schelling model in the context of seating decisions in a classroom.
15 Schelling has two versions of the tipping model. The first version is most thoroughly analyzed in a section of Schelling (1971), in which he refers to it as the “bounded-neighborhood model.” A preview of this model appears in Schelling (1969) and an abridged version is included in Schelling (1978, Chapter 4). The second version of the tipping model is presented in Schelling (1972), focusing solely on the process of neighborhood tipping. Schelling’s diagrammatic treatment of the neighborhood tipping process is slightly different in these two versions of the model, although the central idea is the same.
levels of individuals in each racial group. In a more general sense, as Schelling points out, tipping refers to a process where “something disturbs the original equilibrium” (Schelling, 1971, p. 182). Most interestingly, this “something” may not be “big” in any sense. It is sufficient to tip the system as long as it starts a chain reaction that moves the system further and further away from the original equilibrium situation.16

The tipping model has also generated a great deal of interest.17 Some researchers add a housing market into the model so that neighborhood choices are based on prices instead of preferences as in the original model (Schnare and MacRae, 1978; Becker and Murphy, 2000, Chapter 5). Others try to empirically detect the tipping phenomenon in the transition of racial composition of neighborhoods (Goering, 1978; Clark, 1991; Easterly, 2005; Card et al., 2008a, 2008b).18 In sociology, the tipping idea has been applied to study many other social phenomena and inspired considerable theoretical work on “threshold behaviors” and critical mass models (see, e.g., Granovetter, 1978; Granovetter and Soong, 1983, 1988; Crane, 1991). In economics, there has been some attempt to analyze the tipping phenomenon in a more rigorous setting. Anas (1980) proposes a formal model that explains neighborhood tipping on purely economic grounds without assuming prejudicial preferences as Schelling did. Mobius (2000) and Dokumaci and

16 The concept of tipping has since been widely used by game theorists to refer to the idea that “a small change in the state of a system can cause a large jump in its equilibrium” (Heal and Kunreuther, 2006). The tipping idea has even found its way into the popular culture. Most notably, in a national bestseller, Gladwell (2000) brilliantly explores and illustrates the tipping phenomenon using everyday examples. He very accurately summarizes the three characteristics of tipping dynamics as (1) individual behaviors are interdependent (contagious); (2) accumulation of small changes has significant consequences; and (3) dynamics accelerate once a threshold is reached (Gladwell, 2000, pp. 7-9).

17 The tipping model seems not to have lived up to Schelling’s expectations, which must have been very high. In the prologue to a reprint of his 1971 article “Dynamic Models of Residential Segregation,” Schelling remarked: “I thought the results I got from [the tipping] model were as interesting as those from the checkerboard, but nobody else appeared to think so” (Schelling, 2006, p. 251). Of course, he was comparing the tipping model with the checkerboard model that is probably one of the most well known models in social sciences.

18 The most recent empirical results remain inconclusive. Using census tract level data from 1970 through 2000, Easterly (2005) finds evidence inconsistent with the tipping model. Using the same data but a more sophisticated econometric technique, Card et al. (2008a) find some strong evidence supportive of the tipping model.
Sandholm (2006) follow Schelling more closely and seek to reformulate Schelling’s tipping model in a game-theoretic framework.¹⁹

One important aspect of Schelling’s original tipping model is that the events or activities triggering the tipping process are assumed to be exogenous and thus not generated within the model. In his own words, the initial in-migration of some blacks that tips an all-white neighborhood could be “concerted entry, erroneous entry by a few, organized introduction of a few, redefinition of the neighborhood boundary so that some who were not inside become ‘inside,’ or something of the sort” (Schelling, 1971, p. 182). Because tipping in this model has to be started by some outside events, Mobius (2000) argues that “the basic tipping model suggests no mechanism for moving between an all-white equilibrium and a ghetto equilibrium. The theory can therefore explain the persistence but not the formation of ghettos.” He consequently extends the tipping model by introducing local interaction among residents within the neighborhood that gives an endogenous mechanism of tipping.

Schelling originally presented the checkerboard model and the tipping model independently. Aside from their common subject matter, they are related only in the sense that both models help illustrate the deviation of macrobehavior from micromotives. As already mentioned above, they have distinct formats: one as a simulation and the other as a purely analytical model. The key insight of the checkerboard model—moderate preference for same-color neighbors may lead to complete segregation—is not shown in the tipping model, and the concept of tipping—seemingly unimportant events trigger the movement of a system from one equilibrium to another—is not introduced in the checkerboard model. The tipping model deals with a single neighborhood. It is thus a partial equilibrium model in that although agents (black or white) are assumed to move into and out of the neighborhood, the model does not specify where they come from nor where they go. In contrast, the checkerboard model consists of many

¹⁹ See also Bowles (2003, Chapter 2) for a variation of the bounded-neighborhood model that is used to demonstrate the emergence of “spontaneous order.”
neighborhoods and has a general equilibrium flavor in that agents only move inside the model and their moves generate residential segregation endogenously.  

These differences between Schelling’s checkerboard model and his tipping model have caused research following these two models to develop along different paths that have not overlapped. Similar to Schelling’s simulation with dimes and pennies on a checkerboard, the literature extending his checkerboard model is largely informal, using computer simulation as the primary research tool. The main advantage of this approach lies in the ease of investigating any conceivable variations of the original model. As a consequence, there are now many versions of the checkerboard model, most of which are more complicated than Schelling’s original model. However, the simulation approach also has its disadvantages. Because different simulations are usually written in different programming languages and run on different platforms, simulation models are almost never directly built on earlier work and therefore existing results are only loosely related. In addition, results derived from simulations are less transparent in that it is not always clear which result is driven by which assumptions. These problems can be overcome by developing a mathematical framework to rigorously analyze the checkerboard model. Young (1998) and Zhang (2004a, 2004b) are the early attempts in this direction. This paper is along the same line of research.

The literature inspired by Schelling’s tipping model has been more rigorous. A large fraction of the work in this tradition only borrows Schelling’s idea of tipping to study other social phenomena rather than residential segregation. The studies that examine the dynamics of segregation, like Schelling’s original tipping model, always focus on a single neighborhood (e.g., Mobius 2000; Documaci and Sandholm 2006). However, segregation occurs at the city level and most of the features and consequences of residential segregation have to be understood at the city level. Therefore, in many cases, the multi-neighborhood setting of the checkerboard model is the

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20 The two models also have different neighborhood definitions. In the checkerboard model, neighborhoods are defined with reference to each agent’s residential location. The tipping model focuses on a single neighborhood that has fixed boundaries; an agent is either inside it or outside it.
more appropriate level of analysis than the single-neighborhood setting of the tipping model. This paper represents the first attempt to take the tipping idea developed in a single-neighborhood model and build it into a checkerboard model that can be used to analyze segregation dynamics in a multi-neighborhood area.

I will present a Schelling-type checkerboard model that has the following features: First, it is formulated mathematically as a spatial game and thus can be rigorously analyzed. Second, the concept of tipping is rigorously defined and the trigger of the tipping process is endogenized into the checkerboard model. Third, the tipping of equilibria is crucial for understanding the dynamics of segregation in the checkerboard model. Fourth, the model helps us understand the persistence of segregation in U.S. metro areas.\(^{21}\)

3. A Unified Schelling Model

3.1 Basic setup

This model builds on a Schelling-type checkerboard model first proposed by Zhang (2004a). Zhang studies the model’s long-term dynamics, both analytically and computationally, without introducing the concept of tipping. I will demonstrate here that this model can be extended to incorporate Schelling’s insights on neighborhood tipping. To simplify the analysis, I deviate from Schelling’s original setup by not leaving vacant locations in the residential area. Following Young (1998), I allow individuals to move by switching residential locations. It is as if there exists a centralized agency that processes all the information about who wants to move and which two agents may want to switch.

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\(^{21}\) There is a vast body of literature devoted to the topic of racial housing segregation. Even within the narrow field of theoretical work on the causes of segregation, not all researchers follow the Schelling tradition. In both of Schelling’s models, preference for same-color neighbors is the ultimate cause of segregation. Other researchers have offered alternative explanations based on different factors, including for example, a preference for proximity to locations where another group resides (e.g., Yinger, 1976; Kern, 1981) or racial discrimination in the housing market (e.g., Courant, 1978).
Consider an $N \times N$ lattice graph embedded on a torus, where $N$ is an integer, as the residential area of a city.\footnote{To visualize the spatial structure, one may think of the lattice graph as a two-dimensional grid and the torus as the surface of a doughnut. Note that moving along grid lines on a torus, whether vertically or horizontally, one would never reach a boundary. Wrapping a lattice graph on a torus is a standard simplifying assumption to avoid tedious boundary conditions that would unnecessarily complicate a spatial analysis. Alternatively and equivalently, one could think of the spatial structure as a two-dimensional grid only, but keep in mind that the boundaries are not real because an agent can move beyond the eastern (western) boundary and show up on the western (eastern) boundary or move beyond the northern (southern) boundary and show up on the southern (northern) boundary.} There is a house at each vertex of the graph. Houses are identical in every respect except that some are occupied by black agents and others by white agents. The proportion of blacks in the population is fixed and it has no bearing on the results of the model. A neighborhood is defined locally in the way Schelling did. In particular, an agent takes $2n$ agents around her as her neighbors, where $n$ is an integer much smaller than $N$.

Each agent has a payoff function indicating how much she is willing to pay for a residential location in a particular neighborhood. The payoff function is specified as follows:

$$\omega_i = \beta u(s_i) + \epsilon_i. \quad (1)$$

An agent $i$’s payoff $\omega_i$ has two parts. The first part contains a deterministic term $u$ that denotes $i$’s utility derived from living in a particular neighborhood. This utility is a function of $s_i$, the number of same-color neighbors agent $i$ has. The second part $\epsilon_i$ is a random term. It is assumed to be independent both across agents and across residential locations. Following the evolutionary game theory literature, one could think of this random term as a result of bounded rationality. For example, one may assume that the agent either does not have precise information about the neighborhood racial composition, or is incapable of correctly calculating her payoff, or act impulsively when deciding how much a residential location is worth. Alternatively, one could also interpret the random term as a combination of nonracial characteristics of the neighborhood that residents care about but are unobservable to the modeler. For example, such characteristics may include crime rate, school quality, and proximity to workplace or natural amenities. For the ease of exposition, I will follow the bounded-rationality interpretation in the rest of the paper and
analyze the model as if neighborhood racial composition is the only source of utility. However, it should be noted here that either interpretation gives the same set of results.23

The parameter $\beta$ in the payoff function is a positive constant that determines the relative importance of the random term. If $\beta$ is close to zero, the random term is relatively important and largely determines how much the agent is willing to pay for a particular residential location. As a result, neighborhood racial composition plays a minor role when an agent decides where to reside. If $\beta$ approaches infinity, the random term becomes unimportant and only neighborhood racial composition matters in an agent’s decision. Throughout the paper, I consider the cases with a large $\beta$. It is assumed that every agent has the same $\beta$ and the same utility function $u$.

Assume that every agent prefers to live in an integrated neighborhood. More specifically, every agent attains the highest level of utility when living in a half-black-half-white neighborhood. However, if such evenly mixed neighborhoods are not available, they feel better if they belong to the majority group rather than the minority group. That is, for a white agent, a 30-70 black-white ratio is better than a 70-30 black-white ratio; the opposite is true for a black agent. Survey data suggest that this is a plausible assumption. For example, a recent multi-city study shows that individuals from all racial groups prefer highly integrated neighborhoods, but at the same time they do have a bias in favor of own-group members (Charles, 2001, 2003; Krysan and Farley, 2002). Various reasons could explain the inclination towards one's own race, including cultural concerns, fear of potential hostility from the other group, or dislike of isolation.

To be more specific, I assume that an agent has a single-peaked utility function, depicted in Figure 1. Utility attains its maximum at $n$, which is half of the total number of neighbors. On

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23 The random-utility interpretation, treating $\varepsilon_i$ as a combination of nonracial factors that agents care about, might be more appealing, because it follows standard assumptions in logit models. However, it makes the exposition cumbersome because in my discussion below I would have to repeatedly refer to “moves that increase an agent’s random utility but lower her deterministic utility.” In addition, the random-utility interpretation also causes a minor complication for the welfare analysis in this model. In particular, when discussing social optimality of residential patterns, one needs to consider the realizations of random utilities after agents move. Since I will assume that neighborhood racial composition is the single most important factor that drives locational choices (i.e., let $\beta$ approach infinity), this complication does not affect any of the main results in the model. Therefore, it is better to simplify the exposition and proceed as if the random factors only affect an agent’s willingness to pay but have nothing to do with her utilities.
the left side of $n$, the function is linearly increasing; on the right side of $n$, it is linearly decreasing. It is relatively steeper on the left side, reflecting the assumption that agents would rather belong to the majority group instead of the minority group when half-half mixed neighborhoods are unavailable. Linearity is assumed only for simplicity. Letting $s_i$ be the number of same-color neighbors agent $i$ has, the utility function can be written as:

$$u(s_i) = \begin{cases} \frac{Zs_i}{n}, & s_i \leq n \\ (2Z - M) + \frac{(M - Z)s_i}{n}, & s_i > n \end{cases}$$

where $Z$ is the maximum value of $u$ and $M$ is the value of $u$ when the agent has all neighbors like herself. Because the utility of having no same-color neighbors is normalized to zero, $M$ indicates the difference between living with 100 percent same-color neighbors and no same-color neighbors.

Agents may exchange residential locations. In each period of time, a pair of agents is randomly chosen from two different neighborhoods. The chosen agents are allowed to consider switching residential locations according to their own payoffs. Agents will always trade residential locations when a switch is Pareto-improving (i.e., increases the two agents’ total payoffs). In some cases, both agents have higher payoffs after a switch. In other cases, a switch will lower one agent’s payoff, but it will be carried out as long as the other agent’s gain is enough to compensate the loss. I am therefore assuming that if necessary the two agents can always costlessly negotiate a proper side-payment to make the trade mutually beneficial. This assumption allows me to focus on the sum of the two chosen agents' payoffs because they always attempt to maximize the sum by a joint decision.

If the two agents choose to switch residential locations, the sum of their payoffs after the switch is

$$\omega_i(\text{switch}) + \omega_j(\text{switch}) = [\beta u(s_i|\text{switch}) + \epsilon_i] + [\beta u(s_j|\text{switch}) + \epsilon_j]$$

$$= \beta[u(s_i|\text{switch}) + u(s_j|\text{switch})] + [\epsilon_i + \epsilon_j]$$
= \beta U + \eta,
where \( U = u(s_i|\text{switch}) + u(s_j|\text{switch}) \) and \( \eta = \varepsilon_i + \varepsilon_j \). Similarly, if the two agents do not switch residential locations, the sum of their payoffs is

\[
\omega_i(\cdot|\text{not switch}) + \omega_j(\cdot|\text{not switch}) = [\beta u(s_i|\text{not switch}) + \varepsilon_i'] + [\beta u(s_j|\text{not switch}) + \varepsilon_j']
\]

\[
= \beta [u(s_i|\text{not switch}) + u(s_j|\text{not switch})] + [\varepsilon_i' + \varepsilon_j']
\]

\[
= \beta V + \theta,
\]

where \( V = u(s_i|\text{not switch}) + u(s_j|\text{not switch}) \) and \( \theta = \varepsilon_i' + \varepsilon_j' \).

A switch will happen if and only if it yields higher total payoffs, i.e., \( \beta U + \eta > \beta V + \theta \).

Following McFadden (1973), I assume that \( \eta \) and \( \theta \) are independent and identically follow an extreme value distribution. It is then well known that one can integrate out the random terms of the total payoffs to obtain the probability of switching as:

\[
\Pr(\text{switch}) = \frac{e^{\beta U}}{e^{\beta U} + e^{\beta V}}.
\] (3)

This switch rule is known as a log-linear behavioral rule, which is commonly assumed in the literature (see, e.g., Blume, 1997; Brock and Durlauf, 2001; Young, 1998). It implies that if a switch increases the agents’ utilities, they are more likely to do it. Note that even if a switch decreases these two agents’ utilities, it is still possible that they will do it. The possibility of such a “mistake” depends on \( \beta \). A large \( \beta \) implies that “mistakes” are rarely made. In particular, as \( \beta \) approaches infinity, the probability of a switch approaches 0 if the switch results in lower utilities. In that case, the model reduces to a standard game-theoretic model with agents playing best-reply to their environments. Random mistakes are unobservable. Thus, it is very convenient to integrate them out and analyze the model by working with this behavioral rule.

Define a state \( x \) as an \( N^2 \)-vector, each element labeling a vertex of the \( N \times N \) lattice graph with the color of its occupant. Thus each \( x \) represents a specific residential pattern. Let \( x' \) be the

---

24 For expositional purpose, from this point on, I will refer to any move that lowers the two agents’ utilities (derived from their racial preferences only) as a “mistake.” But I want to note here that a “mistake” does not have to be a real mistake if one interprets the random term in an agent’s payoff function as random utilities.
state at time \( t \), which gives a finite Markov process under the switch rule described in equation (3). Let \( P^\beta \) denote the Markov process (its transition probability matrix). I call \( P^\beta \) a *perturbed process* because agents do not always make “correct” (utility-increasing) decisions depending on the value of \( \beta \). Small values of \( \beta \) imply large perturbations; the perturbation vanishes as \( \beta \) approaches infinity.

Let \( X \) be the set of all possible states of the Markov process. Given a large \( N \), there are a large number of states in \( X \), implying a large number of possible residential patterns. To proceed, I will simplify the analysis by focusing on a single statistic of each residential pattern specified as follows. Given any state, let \( E^D \) be the set of all unordered black-white agent pairs who are neighbors:

\[
E^D = \{(i, j) | \text{agents } i \text{ and } j \text{ are neighbors and have different colors}\}.
\]

A function \( \rho: X \rightarrow \mathbb{N} \) is then defined as the cardinality of the set \( E^D \):

\[
\rho = |E^D|.
\]

For any state \( x' \), \( \rho(x') \) gives the total number of pairs of unlike neighbors. In this model, the value of \( \rho \) (after being properly normalized) serves as a natural index of segregation. It measures the degree of exposure (or potential contact) between the members of the two races; it indicates the extent to which blacks and whites physically confront one another by virtue of sharing a common residential area. A lower \( \rho \) means a higher degree of segregation.\(^{25}\)

The function \( \rho \) also has an attractive property: Its first difference is proportional to the changes in the moving agents’ utilities. This is summarized as the following lemma and its proof is given in the Appendix.

**Lemma 1:** For any two agents \( i \) and \( j \) in two different neighborhoods, the following relation always holds:

\[
\rho(|\text{switch}) - \rho(|\text{not switch}) = -\lambda \{[u(s_i|\text{switch}) + u(s_j|\text{switch})] - [u(s_i|\text{not switch}) + u(s_j|\text{not switch})]\},
\]

where \( \lambda = 2n/M > 0 \) is a constant.

---

\(^{25}\) In empirical analysis, researchers have developed many indices to measure residential segregation. See Massey and Denton (1988) for a detailed discussion.
Equation (4) states that if two agents obtain higher utilities by switching residential locations, their action lowers the value of function $\rho$ by an amount proportional to the increase of their utilities. This equation clearly holds when two agents of the same color exchange residential locations. In that case, their total utilities do not change and the value of function $\rho$ stays the same because the switch does not affect the residential pattern. Equation (4) also holds when a black and a white agent trade residential locations, which is shown in the Appendix. Following the game theory literature, from this point on, I will refer to $\rho$ as the potential function of the spatial game.\textsuperscript{26}

Given that a smaller $\rho$ implies a higher degree of segregation, equation (4) suggests that utility-increasing switches will move the whole system toward segregation. In other words, a mutually beneficial trade of residential locations may have a negative effect on social welfare (sum of all agents’ utilities).\textsuperscript{27} The reason is that the switch of residential locations causes negative externalities. When agents move, they affect the racial composition of both the neighborhoods they leave behind and the neighborhoods they move into. However, they do not take into account such externalities when they decide whether to move. As it will become clear below, equation (4) is crucial for analyzing the dynamics of segregation in this model.

3.2 Tipping

Consider the limiting situation of the Markov process $P^\infty$, i.e., $\beta = \infty$. Remember that an infinitely large $\beta$ implies that agents trade residential locations if and only if such moves increase their utilities (the deterministic terms of their total payoffs). I will refer to this process as the unperturbed process because it is equivalent to setting the random term in an agent’s payoff function to be zero. Define an equilibrium state as one in which there does not exist a pair of

\textsuperscript{26} Equation (4) makes this spatial game a potential game. A game is a potential game if the changes in players’ payoffs can be characterized by the first difference of a function. The function is then called the potential function of the game (Monderer and Shapley, 1996).

\textsuperscript{27} The social welfare implications of residential segregation have long been studied. See, e.g., Kain (1968), Massey and Denton (1993), Borjas (1995), Cutler and Glaeser (1997), and Wasmer and Zenou (2002).
agents who can increase their utilities by trading residential locations. An equilibrium is thus an absorbing state of the unperturbed process because once the system is in it, it will never escape.

It is more convenient to discuss equilibrium states under a specific definition of a neighborhood. From now on, I use the definition of the “Moore neighborhood” in which every agent on the lattice graph considers the eight surrounding agents as neighbors (see Figure 2). Under this definition, a checkerboard residential pattern (each edge of the lattice graph connects a black agent with a white agent) is an equilibrium state. In this case, each agent has exactly four same-color neighbors out of a total of eight neighbors, which gives the highest possible utility level. Therefore, trading residential locations will not increase any agent’s utility and no one has incentive to do it.

In addition to the checkerboard pattern, there exist a large number of equilibria (see Figure 3). For example, a residential pattern with alternating black and white stripes is also an equilibrium state. If a stripe consists of two rows of blacks or whites (panel (b) of Figure 3), a typical agent has five same-color neighbors out of a total of eight. Apparently, no agent is living in a half-half mixed neighborhood and nobody attains the highest level of utility. Nonetheless, it is impossible for any pair of agents to switch residential locations to improve their utilities. The same is true if each stripe consists of more rows of blacks or whites (panel (c) of Figure 3). The most extreme case is a residential pattern in which there is only one stripe of blacks and one stripe of whites (panel (d) of Figure 3). In this case, one observes a complete segregation. But the system is still in equilibrium because no pair of agents can improve their situation by switching residential locations. Therefore, the system has multiple equilibria under the unperturbed process and different equilibria may be associated with different levels of social welfare.

Now come back to the perturbed system that is subject to constant shocks in the form of utility-decreasing moves, i.e., $\beta$ is finite and large. Under this assumption, an equilibrium state as defined above is no longer an absorbing state. Although in an equilibrium state nobody gains utility by moving, there is still a (small) possibility that some agents will move, as implied by the
behavioral rule in equation (3). If agents indeed make such mistaken moves, the system is no longer in equilibrium because now some agents (including the ones that just moved) can improve their utilities by switching residential locations. It is possible that the agents that just moved will reverse their switch thus returning the system to its original equilibrium. However, it is also possible that some other agents will take moves and disturb the system further away from its original equilibrium before it returns, especially now that some of such moves can be utility-improving because of the negative externalities created by other agents’ mistaken moves earlier. In this way, an equilibrium state can be “tipped” away by mistaken moves. The system will then evolve as agents move to increase (and, occasionally, mistakenly decrease) their utilities. At some point, it may reach an equilibrium state that represents a residential pattern very different from the original equilibrium state.

I shall try to illustrate this process with a heuristic example depicted in Figure 4. Define two states as immediately communicating states if the system can travel from one state to the other through a single switch of residential locations between two agents. Also, recall that each state corresponds to a value of the potential function $\rho$, the total number of pairs of unlike neighbors. Consider a series of states lined up on the horizontal axis, each pair of adjacent states being two immediately communicating states. Assume that the value of the potential function at each state is given by the curve plotted in Figure 4. For the time being, assume that the system can only visit these states on the horizontal axis. In particular, if the system is currently in state $x$ (not an end point in Figure 4), only two pairs of agents are allowed to consider a switch. If one pair is picked and they decide to switch residential locations, the system moves to the state on the left of $x$; if the other pair is picked and they decide to switch, the system moves to the state on the right of $x$. In either case, if the picked agents do not switch, the system stays in state $x$.

The potential function shown in Figure 4 has two local minima at states $a$ and $c$. Moving away from $a$ or $c$, whether to the left or to the right, will increase the value of function $\rho$ and
therefore by equation (4) will decrease the moving agents’ utilities. Thus \( a \) and \( c \) are equilibrium states. Any other state is not an equilibrium because it always has an adjacent state with a lower value of \( \rho \), meaning that moving the system to that adjacent state increases the total utilities of the agents who make the switch.

Consider the system that is restricted to the set of moves as shown in Figure 4. What will happen if the system happens to be in state \( a \)? It is very likely that it will stay there because moving in either direction increases the value of \( \rho \) and thus involves utility-decreasing switches that occur with small probabilities. What if the system does move away from \( a \) as a result of a mistaken switch by two agents? It is then very likely that the system will move back instead of moving further away from \( a \), because moving back increases the agents’ utilities and moving further away does the opposite. In fact, starting from any state on the left side of \( b \), it is most likely that the system would soon end up in \( a \) because it only takes a series of utility-improving moves. For the same reason, starting from any state on the right side of \( b \), it is most likely that the system would soon end up in the other equilibrium state \( c \). Therefore, in the long run, the system would spend almost all the time in states \( a \) or \( c \); a visit to any other state will necessarily be transient.

Now the question is whether the system tends to spend more time in \( a \) or in \( c \). Again, suppose the system starts in state \( a \). Notice that it takes a finite number of steps for the system to move to state \( b \), each involving a utility-decreasing move that occurs with a small probability. Given that each of these utility-decreasing moves can happen with a positive probability, there is a positive probability (although extremely small) that this whole series of mistaken moves occur one after another. In that case, the system will move away from \( a \) to the right and reach state \( b \). If it goes beyond \( b \), the system will likely move to state \( c \) quickly because to do so only involves a series of utility-improving switches, each occurring with a probability close to one. Therefore, the system can be tipped away from equilibrium state \( a \) through accumulation of mistaken moves.
and evolve into the other equilibrium $c$. Naturally, state $b$ can be thought of as the *tipping point* because once this point is passed the system tends to quickly travel to $c$ and the chance of returning to $a$ is miniscule.

If the system can be tipped away from equilibrium state $a$, it can also be tipped away from equilibrium state $c$. Tipping away from $c$ will take a series of utility-decreasing switches to move the system from $c$ to $b$, after which it tends to quickly move toward $a$ because moving from $b$ to $a$ only involves utility-increasing switches. As shown in Figure 4, the potential function $\rho$ attains different values at $a$ and $c$. Its value at $c$ is much smaller than its value at $a$. This means that the increase in $\rho$ as the system moves from $a$ to $b$ is much smaller than the increase in $\rho$ as the system moves from $c$ to $b$. In other words, moving the system from $a$ to $b$ requires fewer mistakes (or less serious mistakes measured by the resulting utility losses) than moving the system from $c$ to $b$. Thus, the probability of tipping away from $a$, although very small in absolute value, is many times larger than the probability of tipping away from $c$. Therefore, in the long run, the system would spend most of the time in $c$ or around $c$. The system does visit every state in the long run, but it will not be trapped in any of them because of the positive probability of making utility-decreasing switches. However, it is most difficult to escape from state $c$, not only more difficult than from any of the non-equilibrium states, but also more difficult than from the other equilibrium state $a$.

Here is a quick summary of what I just demonstrated. In the simple example shown in Figure 4, there are two equilibrium states. Tipping could happen to either equilibrium through the accumulation of mistaken switches. However, the chance of tipping is much smaller for one equilibrium than the other, and therefore the equilibrium state that is more resistant to tipping will naturally be observed most of the time in the long run. Interestingly, the equilibrium state more resistant to tipping is also the state that gives the lowest value of the potential function. Recall
that the potential function $\rho$ is defined as the total number of pairs of unlike neighbors. Thus the minimum value of $\rho$ corresponds to the most segregated residential pattern.

In the analysis of the simplified example shown in Figure 4, the movement of the dynamical system is restricted to a small set of states. The unrestricted version of the model is clearly much more complicated. Given a large dimension of the lattice graph, the total number of states (although finite) is very large; even the total number of equilibrium states is fairly large. In addition, each state is directly communicating with a large number of other states because a switch of residential locations between any black agent and any white agent will move the system to a different state. In the rest of this section, I will show that the insight revealed by the simple example in Figure 4 is actually valid in general.

Consider any equilibrium state $x$. A large number of states differ from state $x$ by only one switch because any pair of agents may be chosen and make a switch. By the definition of equilibrium, any state $y$ directly communicating with $x$ will never give the potential function a lower value than $x$ does. That is, the system can move from $y$ to $x$ through a utility-increasing (or utility-preserving) switch. Similarly, there may be some states from which the system can travel to $x$ by two or more switches, each of which either increases or preserves the moving agents’ utilities. Define the *basin of attraction* of an equilibrium state $x$ as the set of all states from which the system can travel to $x$ through a finite number of steps that does not involve a single utility-decreasing switch. The states that belong to two basins of attraction can be thought of as the boundary between them. Starting from any equilibrium state, a series of utility-decreasing switches could move the system away from the equilibrium and reach the boundary of its basin of attraction. Once the system passes the boundary, tipping has occurred because now the system is in a different basin of attraction and some utility-increasing (or utility-preserving) switches will take it to a different equilibrium.
Therefore, I formally define the *tipping* of an equilibrium as the process of moving out of its basin of attraction through the accumulation of low-probability utility-decreasing switches. And a *tipping point* is a point on the boundary of the basin of attraction, beyond which the system can move further away from its original equilibrium with no utility-decreasing switches. Because tipping could happen to any equilibrium of the system, it is useful to identify the ones that are more resistant to tipping than others.

To be intuitively appealing, an equilibrium state $x$ that is resistant to tipping should have the following two properties.

First, if the dynamical system is in state $x$, it has a high probability of remaining in state $x$ in the next period. This implies that a mistaken switch will hardly ever happen when the system is in this equilibrium.

Second, if the dynamical system moves away from $x$ as a result of one or a few mistaken switches, there is a high probability of going back to $x$ within a small number of periods. This implies that even if some mistaken switches have occurred, it takes many more such utility-decreasing moves for the system to get out of the basin of attraction, thus it is likely to return to the original equilibrium before long.\(^{28}\)

If one thinks of “staying in $x$” as returning to $x$ in one period of time, these two features are in fact the same: They both mean that the system, starting from the equilibrium state $x$, will return to $x$ very soon. Therefore, it is sensible to measure an equilibrium state’s resistance to tipping using the expected number of periods for the system to return to the equilibrium given that it starts from the equilibrium.

---

\(^{28}\) One may imagine a real basin and think of tipping as getting a small ball out of the basin along its side wall by constantly shaking the basin. Then the first property simply means that the basin is very steep (and thus it is very difficult to have the ball move away from the bottom); the second property means that the basin is very deep (and thus the ball cannot escape by moving just a few inches away from the bottom and will likely fall back to the bottom again).
Denoting \( r(x) \) as the \textit{resistance to tipping} of equilibrium state \( x \) and \( P^\beta_{xx}(i) \) as the probability of the system moving from state \( x \) to itself in exactly \( i \) periods, I define \( r(x) \) as the following:

\[
r(x) = \frac{1}{\sum_{i=1}^{\infty} iP^\beta_{xx}(i)}.
\]

Given a finite \( \beta \), \( P^\beta_{xx}(1) < 1 \), implying that the sum in the denominator is larger than 1 and therefore \( r(x) < 1 \) for all \( x \). Under this measure, an equilibrium state is considered more resistant to tipping if it takes fewer periods (in expectation) for the system to return to the equilibrium given that it starts there.\(^{29}\)

\[\text{3.3 Main result}\]

I shall summarize the basic setup of the model as follows:

1) A finite number of black and white agents reside on a lattice graph, each occupying a vertex of the graph. 2) Each agent has a preference over the neighborhood racial composition, which is represented by a single-peaked utility function in the form of equation (2) as shown in Figure 1. 3) In each period of time, two agents from different neighborhoods are randomly chosen and they decide on whether to switch residential locations based on the behavioral rule given by equation (3).

Given these assumptions, one can prove the following theorem:

\textbf{Theorem:} Given sufficiently large \( \beta \) and \( t \), the residential pattern of complete segregation is most resistant to tipping and is observed almost all the time.

I will give a sketch of the proof here. First, it is easy to check that the finite Markov chain defined in the model, \( P^\beta \), is irreducible, aperiodic, and recurrent. The process is \textit{irreducible} because all states communicate with each other. It is \textit{aperiodic} because starting in any state \( x \), the

\(^{29}\) This measure of resistance to tipping can also be used to analyze Schelling’s original tipping model. As noted above, Schelling (1971, 1972) assumes that tipping may be triggered by various exogenous events. If one specifies a probability distribution for such events, it is possible to calculate the expected return time given the tolerance levels of residents.
system may enter state $x$ again in any finite period of time. The process is recurrent because starting from any state $x$, the system will reenter state $x$ in the future with probability one. It is a standard result in elementary Markov chain theory that an irreducible, aperiodic, and recurrent Markov chain has a stationary distribution $\mu$ such that $\mu(x)$ is the probability of the system arriving in state $x$ as time goes to infinity, independent of the initial state. $^{30}$ $\mu(x)$ can also be interpreted as the long-run proportion of time that the Markov chain is in state $x$. The following result links an equilibrium state’s resistance to tipping to its stationary probability (see Appendix for the proof).

**Lemma 2:** Let $r(x)$ be the resistance to tipping of state $x$ and $\mu(x)$ its stationary probability, then $\mu(x) = r(x)$ for any state $x$.

Lemma 2 implies that if a state is more resistant to tipping, the dynamical system spends a larger proportion of time in it in the long run. Given the definition of $r(x)$, it means that the long-run proportion of time spent in state $x$ equals the inverse of the expected time between two consecutive visits to $x$.

To complete the proof of the theorem, I only need to show that $\mu(x)$ decreases exponentially with $\rho(x)$. This is given by the following lemma (again, see Appendix for the proof).

**Lemma 3:** If $\mu(x)$ is the stationary distribution of the perturbed process $P^\beta$, then

$$\mu(x) = e^{-\frac{\beta}{\lambda}r(x)+c},$$

where $\beta > 0$, $\lambda = 2n/M > 0$, and $c$ are all constants. $^{31}$

Recall that $2n$ is the size of a neighborhood and $M$ is the utility loss of moving from a neighborhood with 100 percent same-color neighbors to a neighborhood with no same-color neighbors. Given any two states $x$ and $y$, it follows that

$^{30}$ See, for example, Taylor and Karlin (1998, p.247) for a standard reference.

$^{31}$ A similar result was originally derived in statistical mechanics in the analysis of stochastic Ising models (Blume, 1993). Many people, including Schelling himself, have recognized the similarities between the setup of the checkerboard model and the Ising model in physics (Aydınonat, 2005).
\[
\frac{\mu(x)}{\mu(y)} = e^{\frac{\beta M}{2n} [\rho(x) - \rho(y)]}.
\] (7)

Because \(\beta, M,\) and \(n\) are all positive, this ratio is greater than one if and only if \(\rho(x) < \rho(y)\). Thus the most segregated residential patterns, which give the minimum value of \(\rho\), have the largest \(\mu\). This implies that in the long run, such residential patterns are observed more often than any other residential patterns.

Let \(S\) be the set of all states that give a minimum \(\rho\), i.e., \(S = \{x | \rho(x) \leq \rho(y) \text{ for all } y \in X\}\). Suppose \(x\) is a state in \(S\) and \(y\) is an arbitrary state such that \(\rho(y) > \rho(x)\). Given fixed values of \(M\) and \(n\), the ratio \(\frac{\mu(x)}{\mu(y)}\) goes to infinity as \(\beta\) approaches infinity. Therefore, if \(\beta\) is infinitely large, \(\mu(y)\) goes to zero for any \(y\) such that \(\rho(y) > \rho(x)\), and \(\mu(x) > 0\) if and only if \(x\) is in \(S\). In other words, only the states with a minimum \(\rho\) will have a positive probability of being observed in the long run, and only these states have a positive resistance to tipping: \(r(x) = \mu(x) > 0\) if and only if \(x\) is in \(S\). Because \(\rho\) is the total number of unlike neighboring pairs, a state in \(S\) corresponds to complete segregation. Thus, in the long run, complete segregation exists almost all the time given sufficiently large \(\beta\), which establishes the theorem.\(^{32}\)

Equation (7), derived from lemma 3, provides some useful insights into the forces that drive residential segregation. Consider any two states \(x\) and \(y\) such that \(\rho(x) < \rho(y)\), equation (7) implies that the ratio \(\frac{\mu(x)}{\mu(y)}\) increases with \(\beta\) and \(M\). That is to say, complete segregation (with

---

\(^{32}\) In reality, we do not observe “complete segregation” “almost all the time” as predicted by the model. Instead, we only have partial segregation. It is thus worth noting here that the model has abstracted away from many features of the real world for the purpose of highlighting a single mechanism of residential segregation. For this reason, the model does not mimic reality; it only provides a benchmark for thinking about reality, just like many other economic models. One can easily produce partial segregation by introducing more heterogeneous preferences. For example, we can allow a small fraction of agents in this model to have different preferences. Some of them may not care about neighborhood racial composition; others may like rather than dislike neighborhoods with no same-color neighbors; and still others may consider nonracial factors (e.g., local public services or amenities) as more important than racial factors when they choose residential locations. It is not straightforward to extend the analytical framework to incorporate such heterogeneous preferences, although these variations of the model can be easily explored using computer simulations.
minimum \( \rho \) is observed more often if \( \beta \) is larger and if \( M \) is larger. As discussed above, a large \( \beta \) means that the agents’ decision to switch residential locations is primarily determined by their preference over neighborhood racial composition. A large \( M \) means that living with 100 percent same-color neighbors is much better than no same-color neighbors. Therefore, complete segregation in this model is driven by the extent to which racial preference affects residential choice and the utility difference between living in an all-white neighborhood and living in an all-black neighborhood.

More interestingly, equation (7) involves only \( M \) but not \( Z \), the utility of living in a half-black-half-white neighborhood. That is, complete segregation emerges and persists in this model entirely because living with all same-color neighbors is considered better than living with no same-color neighbors (\( M > 0 \)). How much an agent likes the half-half neighborhood (measured by the value of \( Z \)) is totally irrelevant. This is true because once complete segregation emerges, mixed-race neighborhoods are rare and only exist along the color line. For most agents, the choice is between living with all white neighbors and living with all black neighbors. It is thus natural that the difference between these two situations determines the stability of segregation.

Equation (7) also makes it clear why it is necessary to assume a relatively small neighborhood size \( 2n \). Consider an extreme case in which \( 2n = N^2-1 \), i.e., an agent considers all other agents as her neighbors. Then, for any two states \( x \) and \( y \), it is always true that \( \rho(x) = \rho(y) \) and thus \( \frac{\mu(x)}{\mu(y)} = 1 \). That is, any two states are observed with equal probability even if \( \beta \) and \( M \) are large. Therefore, it is a necessary condition for segregation that an agent only cares about the racial composition of nearby agents instead of the whole population.

It is important to note that the analytical techniques developed in this section of the paper are also applicable in other contexts. Consider again the states in \( S \) that minimize the potential function \( \rho \). Lemma 3 implies that
\[
\lim_{t \to \infty, \beta \to \infty} \mu(x) > 0 \text{ if and only if } x \text{ is in } S.
\]

This is precisely the defining property of a \textit{stochastically stable set} in the game theory literature. Then the states in \( S \) are known as the \textit{stochastically stable states} or the \textit{stochastically stable equilibria} of the spatial game (Foster and Young, 1990). Lemma 2, together with lemma 3, also implies that the states in \( S \) are the only states with a positive measure of resistance to tipping in limit. Therefore, in the process of proving the theorem, I have demonstrated that the idea of tipping, as formulated here, is closely related to the concept of stochastically stable equilibrium in evolutionary game theory. In fact, an equilibrium residential pattern’s property of being the most resistant to tipping was shown to be equivalent to being stochastically stable. This theoretical insight that links the intuitive concept of tipping to a broad set of analytical tools in mathematical game theory is likely to be useful for analyzing other types of tipping phenomena.

\section*{4. Simulation}

This section presents agent-based simulations to illustrate the analytical results proved above. I arbitrarily choose a \( 30 \times 30 \) lattice graph, so there are 900 agents in 900 residential locations. Again, the Moore neighborhood definition is used so that every agent considers the eight surrounding agents as neighbors. I parameterize an agent’s utility function by setting \( n = 4, Z = 1, \) and \( M = 0.6. \) Equation (2) is thus reduced to the following mapping:

<table>
<thead>
<tr>
<th>Number of same-color neighbors</th>
<th>0</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
</tr>
</thead>
<tbody>
<tr>
<td>Utility</td>
<td>0</td>
<td>0.25</td>
<td>0.5</td>
<td>0.75</td>
<td>1</td>
<td>0.9</td>
<td>0.8</td>
<td>0.7</td>
<td>0.6</td>
</tr>
</tbody>
</table>

The utility function is invariant to linear transformations and the parameter \( \beta \) always enters the model as a multiplier to the utility. Therefore, the size of \( \beta \) is meaningful only after the unit of the utility function is specified. In the simulations presented below, I set \( \beta = 10. \)
The first simulation starts from the residential pattern labeled as “stripes,” in which two rows of blacks always follow two rows of whites (panel (a) in Figure 5). A typical agent in this initial state has five same-color neighbors. Although this is not the most preferred situation from an agent’s point of view, the two groups are fairly integrated throughout the area. As discussed above, this is an equilibrium state and only utility-decreasing switches can move the system out of the initial state.

Indeed, in a typical run of the simulation, the initial state stays unchanged for some time after the simulation starts. However, this usually does not last long before some mistaken switches occur. And before such mistaken switches are reversed, some other mistaken ones follow. As panel (b) in Figure 5 shows, a series of utility-decreasing switches have been made mistakenly, which has moved the system fairly far away from its original equilibrium. In the initial state, the total number of unlike neighboring pairs is 1260 (the value of “Rho” shown at the bottom of panel (a)). In panel (b), this number has increased to 1359, as a result of all the mistaken switches. By now, the system is out of equilibrium and many pairs of agents can increase their utilities by trading residential locations if they are chosen to consider a switch. Such moves continue to happen over time and the total number of unlike neighboring pairs continues to decline. As panel (c) of Figure 5 shows, the whole area has evolved into a rather segregated residential pattern. In the long run, as shown in panel (d), complete segregation emerges. Every now and then, mistaken moves still occur, which increases the value of \( \rho \), but they only happen occasionally and never push \( \rho \) back too high. Thus complete segregation persists over time.

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33 As indicated in the previous section, the results in the model do not depend on the proportion of the population that is black. In the simulations presented here, the numbers of blacks and whites are chosen to be either roughly the same (as in this first simulation) or exactly the same (as in the next one). This choice is somewhat arbitrary, but it does have one advantage. It clearly shows that the results of the model have nothing to do with the fact that blacks are a minority group in most U.S. cities.

34 Note that the persistence of segregation at the city level is not inconsistent with radical transitions at the neighborhood level. Consider the highly segregated city illustrated in panel (d) of Figure 5. Although the overall segreagational pattern is not changing, the color line can still move over time. As a result, some of
Figure 6 traces the evolution of $\rho$, which helps illustrate how tipping occurred to the original equilibrium. For some time after the simulation starts, the value of $\rho$ does not change. As mentioned above, this is because the starting state is an equilibrium and any switch in that situation will decrease utilities and thus will happen with a very small probability given a fairly large $\beta$. Eventually, such mistaken moves are taken and the value of $\rho$ increases. The accumulation of the mistaken moves soon pushes the system beyond a tipping point, which is shown as the $\rho$ function reaches its maximum in Figure 6. After that point, the $\rho$ function declines rather sharply because many agents find that they can increase their utilities by trading residential locations. However, each such switch actually takes the system one step further toward complete segregation. Occasionally, utility-decreasing switches still happen as indicated by the small increases in $\rho$, but the probability of such events is too small to stop the evolution of the system into complete segregation.

In Figure 6, the small values of the $\rho$ function at the right end correspond to highly segregated residential patterns. In theory, the probability of reversing the process and moving back to the initial state is still positive. However, pushing the $\rho$ function so high requires a large number of utility-decreasing switches. Given that such switches happen with slim chances, the probability that a large number of them happen one after another is virtually zero. Thus, it is almost certain that the tipping of the original equilibrium is irreversible.

The second simulation, shown in Figure 7, starts with a checkerboard pattern. This initial state is not only the most integrated residential pattern but also the socially optimal one because exactly half of every agent’s neighbors are the same color. However, in the long run, it is also tipped into complete segregation. Most interestingly, as shown by the evolution of the $\rho$ function the white neighborhoods close to the color line may change into black neighborhoods and vice versa. Some of these changes at the neighborhood level may even be described as “neighborhood tipping,” which does not lead to drastic changes in the residential pattern at the city level. Therefore, tipping at the city level (as the focus of this paper) is not exactly the same as tipping of a single neighborhood (as studied in Schelling’s original papers and others’ follow-up work); one does not necessarily imply the other.
in Figure 8, this Pareto optimal state is in fact less resistant to tipping than the “stripes”
equilibrium just discussed above. The $\rho$ function in the initial state is at its maximum, and thus
no switches will increase $\rho$. This means that no switches will decrease the moving agents’
utilities. The tipping of this equilibrium state occurs immediately as the $\rho$ function starts to
decrease right away; reversing the process is not likely to happen because it takes a large number
of utility-decreasing switches to do so.

What happened to this socially optimal state is instructive. In the initial state, everybody
lives in a half-half mixed neighborhood, thus it is not a bad idea to switch because after a switch
both agents still end up in a half-half neighborhood. Thus agents are not restrained from trading
residential locations. However, they ignore the externalities they create by their moves. Consider
a black agent in the initial state who decides to trade residential location with a white agent. Both
still have half black neighbors and half white neighbors after the switch. In the neighborhood the
black agent just left, all her neighbors will lose a black neighbor and have one more white
neighbor; each of them now has a lower utility than before. Neither the black agent who is
moving out nor the white agent who is moving in takes these losses of utilities into account.
After their switch, some blacks have 62.5 percent (or 5/8) white neighbors, and some whites have
62.5 percent black neighbors. By assumption, if a black with 62.5 percent white neighbors trades
residential location with a white with 62.5 percent black neighbors, both attain higher utilities.
As a result, they are likely to do it. Unfortunately, as they make the switch, they push both
neighborhoods even further away from the half-half balance. This leaves some of their previous
neighbors even more discontent with their neighborhoods and thus more agents will move. The
integrated residential pattern is soon broken and all the utility-increasing moves will only push
the system even further toward segregation.

Now the question is why a complete segregation cannot be tipped into a socially optimal
residential pattern given that nobody likes segregation. This simulation illustrates the answer.
Once complete segregation emerges, how much agents like integrated neighborhoods becomes irrelevant in their moving decisions because very few locations (along the color line) qualify as such neighborhoods. For example, a black agent living in an all-black neighborhood has essentially two choices when considering a move. One is to trade with another black who lives in another all-black neighborhood. In this case, the switch does not affect the overall residential pattern. The other possibility is to trade with a white agent who lives in an all-white neighborhood. In this case, both agents move from one extreme to the other. As assumed, if integrated neighborhoods are unavailable, agents prefer to live with same-color neighbors. Thus the move is a bad deal for both agents and will be unlikely to occur.

Therefore, although everybody prefers a socially optimal integrated residential pattern, it is not stable. Too many utility-preserving moves can disturb this equilibrium and tip it into segregation. On the other hand, although nobody likes complete segregation, the residential pattern is very stable. Only moving across the color line by a considerable number of agents could disturb the segregation equilibrium, but nobody has incentive to do so because it causes a loss of utility.

This gives one plausible explanation for why segregation is persistent in U.S. metropolitan areas despite the stated preference for integration. Segregation is stable not because people like it, but because any individual who wants to change the situation unilaterally will have to go across the color line, which may not be the desirable thing to do from the individual’s perspective. A similar point was first emphasized by Zhang (2004a).

Some insights from economics can help us better understand the outcome of the model. The failure of the system to escape complete segregation is similar to the phenomenon of “coordination failure” studied by economists in many other contexts. It is the agents’ inability to move simultaneously that make them stuck in a situation nobody likes.35 The fact that the

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35 Many segregational outcomes in Schelling-type models may be understood as coordination failures. See, e.g., Hoff and Sen (2005) and Zhang (2004a).
socially optimal integrated neighborhoods are prone to tipping has to do with the notion of externality.\textsuperscript{36} It is the exclusion of the external costs (imposed on neighbors) in the agents’ decision to move that makes the Pareto optimal residential pattern fall apart. Similarly, it is the exclusion of the external benefits from the agents’ decision to move that keeps complete segregation stable. A perfect analogy to this result in welfare economics is that a competitive market equilibrium fails to produce a Pareto optimal allocation of resources if there exist consumption externalities.\textsuperscript{37}

5. Conclusion

In a series of writings, Schelling (1969, 1971, 1972, and 1978) presents two dynamic models of residential segregation. His checkerboard model uses simulations to show that segregational residential patterns are consistent with various individual preferences, including those with only a moderate inclination to live with same-color neighbors. His neighborhood tipping model demonstrates that seemingly unimportant random shocks could shake a neighborhood out of one equilibrium situation and move it to another equilibrium that is radically different. Using these models, Schelling teaches us that it is difficult to infer “micromotives” from observed “macrobehavior” (Schelling, 1978).

In this paper, I presented a Schelling-type checkerboard model that can be rigorously analyzed. I have shown that segregation emerges and persists in an all-integrationist world where everybody prefers to live in a mixed-race neighborhood, a result consistent with Zhang (2004a) and Pancs and Vriend (2007) but stronger than Schelling’s original finding. Schelling’s idea of tipping, which has always been used in a single-neighborhood setting, is now introduced into the

\textsuperscript{36} Some earlier research on segregation and neighborhood transition also emphasized neighborhood externalities. See, e.g., Bond and Coulson (1989), Borjas (1995), Miyao (1978), and Schnare (1976).

\textsuperscript{37} The situation implies that there is a potential improvement of social welfare through government intervention. One way to achieve it may be to arrange a socially desirable residential pattern and limit individuals’ ability to move afterwards. The constraint on residential choice can take the form of either direct government regulation on certain types of moves or market-based policies such as taxes on or subsidies to certain types of moves. Singapore has a public housing policy with ethnic integration as a primary goal, which was indeed inspired by Schelling’s work (Dodge, 2006, Chapter 17, pp. 141-143).
multi-neighborhood checkerboard model. I have demonstrated that tipping occurs in this checkerboard model in the sense that the accumulation of low-probability random events can perturb an equilibrium residential pattern and trigger the move into a completely different equilibrium residential pattern. The concept of tipping is crucial for understanding the dynamics of segregation in this checkerboard model. Indeed, segregation emerges and persists precisely because tipping tends to occur to integrated residential patterns but not to completely segregated ones. Schelling’s insights, illustrated in his two separate models of segregation, are therefore unified into a single model.

This unified model helps us better understand the persistence of segregation in U.S. metropolitan areas during the past few decades. Following Schelling’s work, it points out the possibility that segregation persists not because people really like it, but because it is a stable state. As the model shows, tipping away from segregation is difficult because the first step involves some blacks moving into predominantly white neighborhoods or some whites moving into predominantly black neighborhoods or both. If people really dislike being isolated in a neighborhood of the opposite color, as survey data seem to suggest, they will not take the first step even though they really like “half black, half white” neighborhoods. Thus, they get stuck with segregation.

This paper is by no means suggesting that segregation will stay forever as a salient feature of urban America. Survey data show that individual residential preferences in the United States have become more favorable toward the general notion of integration (Bobo, 2001). If this trend continues, two conditions of the model may no longer be satisfied in the future. First, the aversion to a neighborhood of the opposite color may fade away (i.e., the asymmetry in the utility function disappears). Second, racial preferences may stop being the primary factors that drive residential location decisions (i.e., parameter $\beta$ is not sufficiently large). In both cases, segregational housing patterns are not stable anymore. Indeed, some recent studies have found that stable integrated neighborhoods, although still rare, are becoming more common (see, e.g.,
Ellen, 2000 and Rawlings et al., 2004), which may be an indication that we are moving in the
direction toward integration.

As a key theoretical contribution of this paper, I have established a link between the often
loosely formulated idea of tipping and the stochastically stable equilibrium, a rigorous
equilibrium concept in evolutionary game theory. I have shown that a residential pattern’s
property of being the most resistant to tipping is equivalent to being a stochastically stable
equilibrium. Because of this theoretical insight, a rich collection of analytical tools in stochastic
evolutionary game theory becomes useful for analyzing tipping phenomena.\footnote{For a comprehensive introduction of stochastic evolutionary game theory, see Young (1998).} Given that
tipping-type dynamics are commonly observed, I anticipate that the method developed in this
paper will find its applications in some other contexts.

This paper can be extended in different directions. First, much analytical work remains
to be completed. To keep the model tractable, I have made a set of rather strong assumptions.
For example, housing market and non-racial neighborhood characteristics are ignored; racial
preferences are symmetric between groups and identical within each group; and moving
opportunities come up independently of agents’ satisfaction or dissatisfaction with their current
neighborhoods. Future research may explore which of these assumptions are crucial for the
analytical results and which can be relaxed.

Second, the model in this paper can also be used to motivate empirical work. For
example, the model predicts that residents’ unwillingness to move across the color line is the
crucial factor to make segregation persistent. One could collect survey data in different cities and
test the theory by correlating individuals’ answers to this question and the degree of segregation
at the city level, similar to what Cutler et al. (1997) and Card et al. (2008a) did in their papers.
The model also predicts that given the asymmetric residential preference, segregation would
persist if parameter $\beta$ is large, i.e., if racial preference is the key factor in driving residential
location choices. This, in principle, can also be tested empirically.
Appendix: Proofs

Proof of Lemma 1: The equation holds trivially when the two agents are of the same color. We only need to deal with the case where the two agents are of different colors.

Consider the following two types of moves: (1) Both agents have less than half same-color neighbors to begin with and thus both will have more than half same-color neighbors after the switch. (2) One agent has more than half same-color neighbors and the other has less than half, and the opposite is true after the switch. These two types of switches and their reverse ones constitute all possible moves in the model and thus it suffices to show that the condition holds in these two cases.

Consider case (1). Suppose agent $A$ has $n-a$ same-color neighbors and agent $B$ has $n-b$ same-color neighbors, $0 \leq a \leq n$ and $0 \leq b \leq n$. After the switch, $A$ will have $n+b$ same-color neighbors and $B$ will have $n+a$ same-color neighbors. The effect on $\rho$ is summarized as follows:

<table>
<thead>
<tr>
<th></th>
<th>Before switch</th>
<th>After switch</th>
<th>Resultant change in $\rho$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$A$’s # of same-color neighbors</td>
<td>$n-a$</td>
<td>$n+b$</td>
<td>$-(a+b)$</td>
</tr>
<tr>
<td>$B$’s # of same-color neighbors</td>
<td>$n-b$</td>
<td>$n+a$</td>
<td>$-(a+b)$</td>
</tr>
</tbody>
</table>

This switch also affects $A$’s and $B$’s neighbors, but because $\rho$ is defined as unordered pairs of unlike neighbors, it will be double counting if one also considers those neighbors. So the total effect on $\rho$ is $-2(a+b)$. That is, the total number of unlike neighboring pairs will decrease by $2(a+b)$ as a result of this switch.

The two agents’ utilities before and after the switch can be summarized as follows:

<table>
<thead>
<tr>
<th></th>
<th>Before switch</th>
<th>After switch</th>
<th>Net gain</th>
</tr>
</thead>
<tbody>
<tr>
<td>$A$’s utility</td>
<td>$Z(n-a)/n$</td>
<td>$(2Z-M)+(M-Z)(n+b)/n$</td>
<td>$[Mb+Z(a-b)]/n$</td>
</tr>
<tr>
<td>$B$’s utility</td>
<td>$Z(n-b)/n$</td>
<td>$(2Z-M)+(M-Z)(n+a)/n$</td>
<td>$[Ma+Z(b-a)]/n$</td>
</tr>
</tbody>
</table>

The total net gain from the switch is thus $(Ma+Mb)/n = -2(a+b)*[-M/(2n)] = \Delta \rho*[-M/(2n)]$, meaning that the total change in the agents’ utilities is proportional to the change in $\rho$.

Consider case (2). Suppose agent $C$ has $n-c$ same-color neighbors and agent $D$ has $n+d$ same-color neighbors, $0 \leq c \leq n$ and $0 \leq d \leq n$. After the switch, $C$ will have $n-d$ same-color neighbors and $D$ will have $n+c$ same-color neighbors. The effect on $\rho$ is summarized as follows:

<table>
<thead>
<tr>
<th></th>
<th>Before switch</th>
<th>After switch</th>
<th>Resultant change in $\rho$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$C$’s # of same-color neighbors</td>
<td>$n-c$</td>
<td>$n-d$</td>
<td>$d-c$</td>
</tr>
<tr>
<td>$D$’s # of same-color neighbors</td>
<td>$n+d$</td>
<td>$n+c$</td>
<td>$d-c$</td>
</tr>
</tbody>
</table>

The total effect on $\rho$ is $2(d-c)$. That is, the total number of pairs of unlike neighbors will change by $2(d-c)$ as a result of this switch.
These two agents’ utilities before and after the switch can be summarized as follows:

<table>
<thead>
<tr>
<th></th>
<th>Before switch</th>
<th>After switch</th>
<th>Net gain</th>
</tr>
</thead>
<tbody>
<tr>
<td>C’s utility</td>
<td>$Z(n-c)/n$</td>
<td>$Z(n-d)/n$</td>
<td>$Z(c-d)/n$</td>
</tr>
<tr>
<td>D’s utility</td>
<td>$(2Z-M)+(M-Z)(n+d)/n$</td>
<td>$(2Z-M)+(M-Z)(n+c)/n$</td>
<td>$(M-Z)(c-d)/n$</td>
</tr>
</tbody>
</table>

The total net gain from the switch is thus $M(c-d)/n = 2(d-c)*[-M/(2n)] = \Delta \rho *[-M/(2n)]$, still proportional to the change in $\rho$, and the multiplier of $\Delta \rho$ is the same $-M/(2n)$.

Therefore we have established that $\rho([\text{switch}]) - \rho([\text{not switch}]) = \lambda \{[u(s_i|\text{switch})] - [u(s_i|\text{not switch})] + u(s_j|\text{not switch})]\}$, where $\lambda = 2n/M > 0$.

Q.E.D.

Proof of Lemma 2: For any two states $x$ and $y$, let $P_{xy}$ be the transition probability from state $y$ to state $x$ and $m_{xy}$ the expected number of time periods for the system, starting from state $y$, to arrive at state $x$. Starting from state $y$, let’s assume 1 period has passed and examine the expected number of additional time periods it takes for the system to arrive at state $x$. With probability $P_{xy}$ the system is already in $x$ and it takes 0 additional time periods. With probability $P_{ky}$ the system is in state $k \neq x$, and it takes $m_{ky}$ periods to arrive at $x$. Therefore, $m_{xy}$ can be written as

$$m_{xy} = 1 + P_{xy} \cdot 0 + \sum_{k \neq x} P_{yk} m_{ky} = 1 + \sum_{k \neq x} P_{yk} m_{ky}.$$  

Multiply both sides of the equation by $\mu(y)$ and sum over all states to obtain

$$\sum_y \mu(y)m_{xy} = \sum_y \mu(y) + \sum_{k \neq x} \sum_y \mu(y)P_{yk} m_{ky} = 1 + \sum_{k \neq x} m_{ky} \mu(k).$$

This implies that $1 = \sum_y \mu(y)m_{xy} - \sum_{k \neq x} \mu(k)m_{ky} = \mu(x)m_{xx}$. It immediately follows that $\mu(x) = 1/m_{xx} = r(x)$.

Q.E.D.

Proof of Lemma 3: This result is a special case of theorem 6.1 in Young (1998). Let $X$ be the set of all states and define a function $\pi : X \rightarrow [0,1]$ as follows

$$\pi(x) = \frac{e^{-\frac{\beta}{\lambda} \rho(x)}}{\sum_{z \in X} e^{-\frac{\beta}{\lambda} \rho(z)}}.$$  

Let $P_{xy}$ be the transition probability from state $x$ to $y$. It is straightforward to check that $\pi(x)$ satisfy the detailed balance condition $\pi(x)P_{xy} = \pi(y)P_{yx}$. If $x = y$ or $P_{xy} = P_{yx} = 0$, function $\pi$ trivially satisfies the detailed balance condition. If $x \neq y$ and $P_{xy} \neq 0$ or $P_{yx} \neq 0$, then it must be true that $x$ and $y$ differ only at two locations. In the case the agents at these two locations are chosen, a switch between them will change the state from one to the other. There are a total of $N^2$ agents on the lattice graph, therefore these two agents will be chosen with probability $1/[N^2(N^2-2n-1)]$. (It is assumed that the two agents chosen are not neighbors, and thus once one is chosen, its $2n$ neighbors won’t be considered.) Let $\gamma = 1/[N^2(N^2-2n-1)]$, it follows that
\[
\pi(x)P_{xy} = \frac{e^{\frac{\beta}{\lambda}p(x)}}{\sum_{z \in X} e^{\frac{\beta}{\lambda}p(z)}} \left\{ y \cdot \frac{e^{\beta u} + e^{-\beta v}}{e^{\beta u} + e^{-\beta v}} \right\}
\]
\[
= \frac{e^{-\frac{\beta}{\lambda}p(x)}}{\sum_{z \in X} e^{-\frac{\beta}{\lambda}p(z)}} \left\{ y \cdot \frac{e^{\beta[u(s_j, \text{switch})+u(s_j, \text{switch})]+\beta[u(s_j, \text{switch})+u(s_j, \text{not switch})]+\beta[u(s_j, \text{not switch})+u(s_j, \text{not switch})]}}{e^{\beta u} + e^{-\beta v}} \right\}
\]
\[
= \frac{e^{-\frac{\beta}{\lambda}p(x)}}{\sum_{z \in X} e^{-\frac{\beta}{\lambda}p(z)}} \left\{ y \cdot \frac{e^{\beta[u(s_j, \text{not switch})+u(s_j, \text{not switch})]+\beta[p(s_j, \text{switch})]+\beta(p(s_j, \text{switch})+\beta(p(s_j, \text{not switch})]}}{e^{\beta u} + e^{-\beta v}} \right\}
\]
\[
= \frac{e^{-\frac{\beta}{\lambda}p(x)}}{\sum_{z \in X} e^{-\frac{\beta}{\lambda}p(z)}} \left\{ y \cdot \frac{e^{\beta[u(s_j, \text{not switch})+u(s_j, \text{not switch})]+\beta[p(s_j, \text{not switch})]}}{e^{\beta u} + e^{-\beta v}} \right\}
\]
\[
= \frac{e^{-\frac{\beta}{\lambda}p(x)}}{\sum_{z \in X} e^{-\frac{\beta}{\lambda}p(z)}} \left\{ y \cdot \frac{e^{\beta y} + e^{-\beta y}}{e^{\beta u} + e^{-\beta v}} \right\} = \pi(y)P_{yx}.
\]

Therefore, \(\sum_{x \in X} \pi(x)P_{xy} = \sum_{x \in X} \pi(y)P_{yx} = \pi(y) \sum_{x \in X} P_{yx} = \pi(y) \cdot 1 = \pi(y)\), which means that function \(\pi\) is a stationary distribution of the perturbed process. Because the Markov process is finite and irreducible, it has a unique stationary distribution. Given that \(\mu\) is a stationary distribution of the process, it must be true that \(\mu(x) = \pi(x)\) for any state \(x\). It follows that
\[
\mu(x) = \pi(x) = e^{\frac{\beta}{\lambda}p(x) - \sum_{z \in X} \frac{\beta}{\lambda}p(z)} = e^{\frac{\beta}{\lambda}p(x) + c},
\]
where \(c = -\ln \sum_{z \in X} e^{\frac{\beta}{\lambda}p(z)}\) is a constant. This implies that states with minimum \(\rho\) have the highest probability of being visited. In the long run, with a \(\beta\) close to infinity, such states are observed almost all of the time.

\(Q.E.D.\)
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Figure 1: An agent’s utility as a function of neighborhood racial composition

\[ u(s_i) = \frac{(2Z-M) + (M-2)Z}{n} \]

\[ u(s_i) = \frac{Zs_i}{n} \]

# of same-color neighbors \((s_i)\)

Figure 2: Moore neighborhood
Figure 3: Equilibrium residential patterns
Figure 4: Tipping between two equilibrium states in a restricted process
Figure 5: Tipping of an equilibrium residential pattern ("stripes")
Figure 6: Evolution of the potential function, starting with an equilibrium residential pattern ("stripes")
Figure 7: Tipping of the socially optimal residential pattern

(a) 

(b) 

(c) 

(d)
Figure 8: Evolution of the potential function, starting with the socially optimal residential pattern