Equilibria of culture contact derived from ingroup and outgroup attitudes

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Abstract

Modern societies feature an increasing contact between cultures, yet we have a poor understanding of what the outcomes might be. Here we consider a mathematical model of contact between social groups, grounded in social psychology and analyzed using tools from statistical physics. We use the model to study how a culture might be affected by immigration. We find that in some cases residents’ culture is relatively unchanged, but in other cases residents may adopt the opinions and beliefs of immigrants. The decisive factors are each group’s cultural legacy and its attitudes towards in- and out-groups. The model can also predict how social policies may influence the outcome of culture contact.

1 Introduction

Contact between cultures is a prominent feature of modern society, driven by large-scale migration, global media, communication networks, and other socio-economical forces (1). Understanding how human cultures interact is crucial to such issues as immigration management and the survival of national and minority cultures (2, 3), yet the dynamics of culture contact are poorly known. Here we explore the problem considering a single cultural trait that can take two forms, such as being in favor or against the death penalty, or whether to wear or not a particular piece of clothing. We are interested in: 1) how the two trait forms are distributed among subgroups in a population, e.g., residents and immigrants, males and females, social classes, etc.; and 2) how different subgroups influence each other’s traits. We start with assumptions about how individuals may change their opinions and behaviors
as a consequence of social interactions, and then derive the population-level consequences of such assumptions using tools from statistical physics. We study the case of immigration in detail, and conclude that culture contact may sometime result in residents taking on the opinions and beliefs of immigrants, depending on each group’s cultural legacy, its attitudes toward in- and out-groups, and social policies.

Our main assumption about how individuals may change their traits is the so-called “similarity-attraction” hypothesis of social psychology: an individual tends to agree with those who are perceived as similar to oneself, and to disagree with those who are perceived as different (4, 5, 6, 7). Additionally, individuals can be influenced by other forces favoring one form of a trait over the other. For instance, a media campaign may advertise in favor or against a given idea or behavior. To determine the distribution of opinions among subgroups of a population (after they have been in contact for some time) we apply statistical mechanics, a branch of theoretical physics that studies the collective properties of systems composed of many parts that interact according to given rules. These techniques were originally developed to derive the laws of thermodynamics from molecular dynamics (8), but have been applied also to biological (9, 14) and social systems, including models of social choice (10, 11, 12, 13, 15). The latter consider how collective opinions or choices emerge within a homogeneous social group. The model we discuss here is, to our knowledge, the first one to consider populations consisting of different social groups in interaction.


2 General framework

We now outline a general framework for modelling group-level consequences of interactions among individuals (16). Individual \( i \) is described by a binary variable \( s_i = \pm 1 \), representing the two possible forms of the considered trait (\( i = 1, \ldots, N \)). A group is characterized by the mean value \( m = \frac{1}{N} \sum_i s_i \), which can be measured by, say, a referendum vote or a survey. To apply statistical mechanics we formalize as follows the interaction between individuals. Let \( L_{ik} \) be the similarity that individual \( i \) perceives with \( k \) and assume that \( L_0 \) is the level of similarity above which individuals tend to agree, and below which they tend to disagree. We can then recast the similarity-attraction hypothesis in the form of a minimization rule, assuming that \( i \), when interacting with \( k \), tends to assume the trait value that minimizes the quantity

\[
H_{ik}(s_i, s_k) = -(L_{ik} - L_0)s_is_k
\]  

This assumption agrees with the similarity-attraction hypothesis because, when \( L_{ik} > L_0 \) the expression is minimized by agreement \((s_is_k = 1)\), and when \( L_{ik} < L_0 \) by disagreement \((s_is_k = -1)\). For simplicity we let \( L_{ik} - L_0 = J_{ik} \) in the following. Then \( J_{ik} > 0 \) favors agreement and \( J_{ik} < 0 \) favors disagreement. We say “favor” because we do not assume strict minimization. Rather, we assume that a trait value yielding a lower value of \((1)\) occurs with a higher probability, but not with certainty (see appendix). The rationale for this assumption is that similarity to others is not the sole determinant of an individual’s trait (6, 7).
When an individual interacts with many others, we assume that she tends to minimize the sum, $H_i$, of all functions (1) relative to each interaction:

$$H_i(s_i) = \sum_k H_{ik}(s_i, s_k) = - \sum_k J_{ik}s_is_k$$

(2)

where the sum extends over all individuals with whom $i$ interacts. In summary, the effect of changing one’s trait is gauged according to whether it makes an individual agree or disagree with others.

The effect of factors such as the media or social norms is, for each individual, to favor a particular trait value. This can be included in our minimization scheme by adding a term to (2):

$$H_i(s_i) = - \sum_k J_{ik}s_is_k - h_is_i$$

(3)

The added term means that $s_i = 1$ is favored if $h_i > 0$, while $h_i < 0$ favors $s_i = -1$. We now define a group-level function $H$ as the sum of individual functions:

$$H(s) = \sum_i H_i(s_i) = - \sum_{i,k} J_{ik}s_is_k - \sum_i h_is_i$$

(4)

where $s = \{s_1, \ldots, s_N\}$ is the set of all individual traits. The function $H$ is referred to as the system Hamiltonian in statistical mechanics, where it usually arises from consideration of a system’s physical properties. Here, on the other hand, we have designed the function $H$ so that lower values of $H$ correspond to group states that, given our assumptions about individual psychology, are more likely to occur. It is this property that allows us to use statistical mechanics (8). Note that we do not assume that individuals explicitly carry out,
or are aware of, the computations in equation (4).

3 Culture contact in immigration

We consider a large and a small group, referred to, respectively, as residents \( R \) and immigrants \( I \). We are interested in the effect of one group upon the other, i.e., how culture contact changes the mean trait values in the two groups. If residents and immigrants have markedly different culture, the similarity-attraction hypothesis implies that \( J_{ik} \) should be positive for interactions within a group and negative for interactions between groups. A simple assumption (called mean field in statistical mechanics) is that \( J_{ik} \) depends only on group membership. This corresponds to the ingroup and outgroup concepts of social psychology (17) and can be formalized as follows:

\[
J_{ik} = \begin{cases} 
\frac{J_{res}}{2N} > 0 & i,k \in R \\
\frac{J_{int}}{2N} < 0 & i \in R, k \in I, \text{ or } i \in I, k \in R \\
\frac{J_{imm}}{2N} > 0 & i,k \in I 
\end{cases}
\]  

(5)

where the factor \( 1/2N \) guarantees that the group function, equation (4), grows proportionally to the number of individuals. Before the two groups interact, residents and immigrants are each characterized by a cultural legacy that includes given mean values of the considered trait, say \( m^*_{res} \) and \( m^*_{imm} \). Our goal is to predict the values \( m_{res} \) and \( m_{imm} \) after the
interaction has taken place. To describe the effect of cultural legacies we reason as follows. Within a group, a mean trait \( m^* \neq 0 \) means that the two forms of the trait are not equally common. Thus preexisting culture can be seen as a force that favors one trait value over the other, and can be modeled by a force term as in (3). In other words, a model in which individuals are biased so that the mean trait is \( m^* \) is equivalent to a model with unbiased individuals subject to a force \( h^* \) of appropriate intensity. The latter can be calculated by standard methods of statistical mechanics as

\[
h^* = \text{atanh}(m^*) - Jm^*
\]  

(6)

where \( \text{atanh}(. \cdot) \) is the inverse hyperbolic tangent and \( J \) is the ingroup attitude (8, 16). Statistical mechanics also allows to calculate the values of \( m_{\text{res}} \) and \( m_{\text{imm}} \) after culture contact as the solution of a system of equations (see appendix):

\[
\begin{align*}
m_{\text{res}} &= \tanh \left( (1 - \alpha)J_{\text{res}}m_{\text{res}} + \alpha J_{\text{int}}m_{\text{imm}} + h^*_{\text{res}} \right) \\
m_{\text{imm}} &= \tanh \left( (1 - \alpha)J_{\text{int}}m_{\text{res}} + \alpha J_{\text{imm}}m_{\text{imm}} + h^*_{\text{imm}} \right)
\end{align*}
\]  

(7)

where \( \tanh(\cdot) \) is the hyperbolic tangent, \( \alpha \) is the fraction of immigrants in the compound group, and \( h^*_{\text{res}} \) and \( h^*_{\text{imm}} \) are calculated for each group according to (6). Numerical analysis of these equations reveals two main patterns of behavior depending on the sign of the product \( m^*_{\text{res}}m^*_{\text{imm}}J_{\text{int}} \) (figure [1]). When \( m^*_{\text{res}}m^*_{\text{imm}}J_{\text{int}} > 0 \) a small fraction of immigrants causes only small changes in residents’ trait, as intuition would suggest. This includes two distinct cases: either the two groups agreed before the interaction \( (m^*_{\text{res}}m^*_{\text{imm}} > 0) \) and have
Figure 1: Influence of immigration on resident culture. Each panel portrays the mean resident trait after culture contact, $m_{\text{res}}$, as a function of the fraction of immigrants, $\alpha$, and the strength and sign of outgroup attitude, $J_{\text{int}}$. Dramatic shifts in resident trait occur only when $J_{\text{int}}m^*_{\text{res}}m^*_{\text{imm}} < 0$ (b, d) but not when $J_{\text{int}}m^*_{\text{res}}m^*_{\text{imm}} > 0$ (a, c), where $m^*_{\text{res}}$ and $m^*_{\text{imm}}$ are, respectively, residents’ and immigrants’ mean traits before cultural interaction. Parameter values: $J_{\text{res}} = 1$, $J_{\text{imm}} = 0.7$, $m^*_{\text{res}} = 0.7$, $m^*_{\text{imm}} = -0.5$ (a, c) or $m^*_{\text{imm}} = 0.5$ (b, d).
Figure 2: Ingroup attitudes and shifts in cultural traits. Each line shows the critical fraction of immigrants, \( \alpha_c \), above which a sudden shift in residents’ trait is observed, as a function of the strength of residents’ ingroup attitude, \( J_{res} \). The curves are identical for the cases in figure 1b and c, i.e., for positive or negative \( J_{int} \). For each value of \( m^*_{res} \) (different lines), lowering \( J_{res} \) sharply increases the fraction of immigrants that can be sustained before residents significantly change trait.

similar culture (\( J_{int} > 0 \), figure 1b) or they disagree and have dissimilar culture (\( m^*_{res} m^*_{imm} < 0 \) and \( J_{int} < 0 \), figure 1d). The second pattern of results occurs when \( m^*_{res} m^*_{imm} J_{int} < 0 \), in which case there exists a critical fraction of immigrants, \( \alpha_c \), above which residents suddenly change to a nearly opposite mean trait value. This can happen either when the two groups agree and have dissimilar culture (\( m^*_{res} m^*_{imm} > 0 \) and \( J_{int} < 0 \), figure 1b) or when the groups disagree and have similar culture (\( m^*_{res} m^*_{imm} < 0 \) and \( J_{int} > 0 \), figure 1c). This dramatic phenomenon only exists when attitudes toward the outgroup (either positive, figure 1b, or negative, figure 1c) are strong enough. The shift can thus be inhibited by decreasing the magnitude of \( J_{int} \). According to our assumptions, this amounts to making the groups less similar when they are similar and less different when they are different.
Figure 3: Social incentives and shifts in cultural traits. Both panels plot the critical fraction of immigrants, $\alpha_c$, above which a sudden shift in residents’ trait is observed, as a function of the strength of a social incentive capable of affecting individual traits, modeled as an external force, $h$ (equation 3). 

**a**: A social incentive is added to the situation in figure 1b ($h = 0$, red dot in this figure). If $h$ favors (opposes) the residents’ original trait value, the critical fraction of immigrants is raised (lowered).

**b**: A social incentive is added to the situation in figure 1d, in which no dramatic change is predicted. When $h$ is decreased from 0 toward negative values, residents’ trait changes very slightly (not shown), until a critical value $h_c$ is reached when a sudden shift in residents’ trait is predicted with a very small fraction of immigrants. Parameter values as in figure 1.
There are other ways in which dramatic shifts in residents’ trait may be prevented. It is possible, for instance, to reduce the strength of residents’ ingroup attitude, $J_{\text{res}}$ (figure 2). According to the similarity-attraction hypothesis, a higher value of $J_{\text{res}}$ corresponds to higher ingroup similarity. Hence our model suggests that a more culturally homogeneous groups has a greater risk of undergoing a dramatic transition when confronted with an immigrant culture. That is, encouraging cultural diversity among residents may make their culture more robust. One may also try to influence individuals directly, introducing, e.g., social incentives that encourage a given trait (modeled by $h$ terms as in equation 3).

Consider a situation in which residents are predicted to change trait when the fraction of immigrants passes a critical value, $\alpha_c$ (figure 1b). In such a case $\alpha_c$ can be increased subjecting individuals to an incentive $h$ that favors the residents’ original trait (figure 3a). An incentive in the opposite direction, on the other hand, decreases $\alpha_c$, suggesting that the impact of social policies can be dramatic. Indeed, we show in figure 3 that an incentive $h$ can trigger dramatic changes even when these are impossible with $h = 0$, as for instance in figure 1d.

## 4 Discussion

Clearly, our model is only an approximation to the complexities of culture contact. Yet it exhibits a rich set of behaviors that, we hope, may help to understand this complex phenomenon and promote the development of more refined models. Our approach can be developed by considering more individual traits (so that changes in culture as a whole can be assessed), more realistic patterns of interactions among individuals (social networks), indi-
individual variability in ingroup and outgroup attitudes, and more complex rules of individual interaction. Our basic hypothesis (similar individuals imitate each other more strongly than dissimilar ones), however, is well rooted in social psychology (4, 6), including studies of intergroup behavior (5, 7), and we expect it to remain an important feature of future models.

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References

[1] Lull, J. Media, communication, culture. Polity Press, Cambridge, UK, (2000).

[2] Cornelius, W. A., Martin, P. L., and Hollifield, J. F., editors. Controlling Immigration: A Global Perspective. Stanford University Press, Stanford, CA, (1994).

[3] Givens, T. and Luedtke, A. Comp. Europ. Politics 3, 1–22 (2005).

[4] Byrne, D. The attraction paradigm. Academic Press, New York, (1971).

[5] Grant, P. R. Can. J. Behav. Science 25, 28–44 (1993).

[6] Byrne, D. J. Pers. Soc. Psychol. 14, 417–431 (1997).
[7] Michinov, E. and Monteil, J.-M. *Europe. J. Soc. Psychol.* **32**, 485–500 (2002).

[8] Thompson, C. *Mathematical Statistical Mechanics*. Princeton University Press, Princeton, NJ, (1979).

[9] Amit, D. *Modeling brain function*. Cambridge University Press, Cambridge, (1989).

[10] Schelling, Thomas C. *Hockey Helmets, Concealed Weapons, and Daylight Saving: A Study of Binary Choices with Externalities*. *Journal of Conflict Resolution* **32**, 381–428 (1973).

[11] Granovetter, M. S. *Threshold models of collective behavior*. *American Journal of Sociology* **83**, 1420-1443 (1978).

[12] Durlauf, S. N. *Proc. Natl. Acad. Sci. U.S.A.* **96**, 10582–10584 (1999).

[13] Watts, Duncan J. *A simple model of information cascades on random networks*. *Proceedings of the National Academy of Science, U.S.A* **99**, 5766-5771 (2002).

[14] Arbib, M. A. *The Handbook of Brain Theory and Neural Networks*. MIT Press, 2 edition, (2003).

[15] Macy, M. W., Kitts, J. A., Flache, A., and Benard, S. In *Dynamic Social Network Modeling and Analysis*, 162–173. National Academic Press, (2003).

[16] Contucci, P. and Ghirlanda, S. *Quality & Quantity* (in press).

[17] Brown, R. *Group processes: dynamics within and between groups*. Blackwell Academic Publishing, Oxford, 2 edition, (2000).
Appendix

Statistical mechanics assigns to each system configuration $s$ a probability that is inversely related to the value of $H(s)$ through an exponential function (8)

$$Pr(s) = \frac{e^{-H(s)}}{\sum_s e^{-H(s)}}$$

where the denominator is a normalization factor ensuring $\sum_s Pr(s) = 1$. Equation (8) can be proved in a number of important cases (8), but is also used heuristically in more general settings (18). In particular, the exponential is the only function compatible with basic assumptions relevant for social as well as physical systems, such that the probability of events relative to two independent subsystems must multiply while other quantities (e.g., entropy, see below) must add (19).

The probability distribution (8) is used to calculate the so-called free energy, from which one can derive system-level quantities such as the mean trait values $m_{res}$ and $m_{imm}$. The free energy is defined as the difference between internal energy $U$, i.e. the average of
$H$ with respect to (8), and entropy $S = - \sum_s Pr s \ln Pr s$:

$$F = U - S$$  \hfill (9)

For the model in the main text one may show that (20)

$$U = - \frac{1}{2} (J_{\text{res}} (1 - \alpha)^2 m_{\text{res}}^2 + J_{\text{imm}} \alpha^2 m_{\text{imm}}^2 + 2 J_{\text{int}} \alpha (1 - \alpha) m_{\text{imm}} m_{\text{res}}) +$$

$$- (1 - \alpha) h_{\text{res}}^* m_{\text{res}} - \alpha h_{\text{imm}}^* m_{\text{imm}}.$$  \hfill (10)

and

$$S = (1 - \alpha) \left( - \frac{1 + m_{\text{res}}}{2} \ln \left( \frac{1 + m_{\text{res}}}{2} \right) - \frac{1 - m_{\text{res}}}{2} \ln \left( \frac{1 - m_{\text{res}}}{2} \right) \right) +$$

$$\alpha \left( - \frac{1 + m_{\text{imm}}}{2} \ln \left( \frac{1 + m_{\text{imm}}}{2} \right) - \frac{1 - m_{\text{imm}}}{2} \ln \left( \frac{1 - m_{\text{imm}}}{2} \right) \right).$$  \hfill (11)

The values of $m_{\text{res}}$ and $m_{\text{imm}}$ are obtained from these expressions by minimizing (9), which yields equation (7).