

# International Dimensions of Civil War<sup>1</sup>

Kristian Skrede Gleditsch<sup>2</sup>

22nd March 2002

<sup>1</sup>Prepared for the Annual Meeting of the International Studies Association in New Orleans, LA, 24-27 March 2002. This is a very early draft in constant revision. Please email me for the most recent copy. This research was supported by the Human Security Project at Harvard University and a travel grant from the University of California. I am grateful to Kyle Beardsley and Ismene Gizelis for helpful comments

<sup>2</sup>Department of Political Science, University of California, San Diego, La Jolla, California, USA, 92093-0521, email: [kgleditsch@ucsd.edu](mailto:kgleditsch@ucsd.edu)

## **Abstract**

Most existing research has related civil war to various factors processes within individual states. Many cases of civil wars, however, appear to have a transnational character, where the risk of conflict in one state is influenced by wars in other states as well as actors and events spanning national boundaries. In this paper, I examine how interaction and processes between states influence the likelihood of conflict within states. I evaluate a series of hypothesis on how transnational contagion and regional factors may influence both the risk of conflict and the prospect for maintaining peace. I evaluate these external influences in a model with various country specific factors often associated with civil wars. The empirical findings indicate that many such linkages between states and regional factors strongly influence the risk of conflict. Some of the commonly inferred effects of attributes of individual states on conflict appear to change once linkages between states are considered. The proposed model can successfully classify conflicts in the estimation sample, and displays good predictive ability in an out of sample forecast.

## 1 Introduction

Many of the contemporary conflicts in the international system involve combinations of formally recognized states and less clearcut types on non-state actors. Consider, for example, the conflict in Bosnia. This was nominally a civil war between ethnic Serbs and Croats and the central government of the newly independent Republic of Bosnia-Herzegovina. The conflict nonetheless had a clear transnational character, as the Serb and Croat units fighting in Bosnia received various kinds of support from ethnic kin in the remaining Republic of Yugoslavia as well as the Republic of Croatia. The conflict emerged over concerns among the different ethnic groups on the status of an independent Bosnia and their desires to secede or to be unified with neighboring national states. The influence of actors outside the boundaries of Bosnia itself strongly influenced outcomes on the battlefield and the relative powers of the actors.

Researchers have tended to draw a sharp distinction between conflict among states and conflict within states, and the two types of conflict have generally been studied in separation of one another. Cases such as Bosnia, however, attest to the problems involved in imposing a strict separation between “civil” wars within states and interstate wars. The international implications of such transnational conflicts highlight how research on international violence cannot ignore wars where the two main antagonists are not nation states. Similarly, studies of civil war cannot limit themselves exclusively to factors contained within individual states, but must consider how influences from other states alter the likelihood that antagonistic groups will resort to violence.

In this paper, I examine how interaction and processes between states influence the likelihood of conflict within states. Consistent with arguments about transnational contagion between states, I find that the presence of conflict in other states strongly influences the risk of civil war. I examine a number of linkages between states that have been hypothesized to increase both the risk of conflict and prospect for peace, and provide empirical support for that many of such linkages between states alter the risk of conflict. Finally, many of the commonly inferred effects of attributes of individual states appear to change once we include linkages between states in an empirical model of conflict.

## 2 Civil and transnational wars

The field of international relations has tended to focus almost exclusively on interstate conflict or disputes taking place among parties that are formally sovereign states. That states constitute the principal actors has been taken as somewhat of an article of faith in the field of international relations. Relations between states are sometimes held to be fundamentally different from relations within formally sovereign states due to the condition of anarchy and absence of formal authority in the international system. In this perspective, it is sometimes simply assumed that sovereignty is effective within nations, thereby making civil war qualitatively different from interstate conflict. Yet, sovereignty is obviously less than fully effective within many existing states, and similar problems of enforcement and contracting under anarchy could obtain within states as well. When relating “Warre” to anarchy, Hobbes had relations within states in mind, not the international system.

Empirically, conflict between states is in many ways a relatively limited share of the

armed conflict involving nation states in the contemporary international system. According to one account (Wallensteen and Sollenberg 1999), only six out of 103 armed conflicts in the period 1989-97 were interstate conflicts (see also Holsti 1996: 22). Many argue that the ratio of “international” to “civil” conflict has been increasing over the 20th century (Pfetsch and Rohloff 2000). Even though the reliability of such conflict proportions over time is questionable - coding biases and media exposure probably understate the amount of civil war and conflict outside the Central European state system in earlier time periods - this result still seems to run counter to what one would expect given the large increase in sovereign nations states over the period. More states should increase the opportunities for interstate wars that previously would have been considered “civil wars.”

Although it is possible that internal and external conflicts are analytically distinct, it is in practice often problematic to restrict analyses to conflicts classified as interstate wars only. Upon closer scrutiny, most lists of interstate wars appear to cover only a subset of what observers typically perceive as armed conflict and war, or the phenomenon researchers seek to make some generalizations about. To actually determine whether a given conflict is “interstate” or not is often less straightforward than might be assumed.

The Correlates of War has tried to classify civil and interstate conflict as mutually exclusive categories (provides the most recent discussion of the COW war data Sarkees 2000). This distinction is in practice often difficult to apply when classifying conflicts. Some conflicts have switched categories between updates of the COW war data.<sup>1</sup> Similar ambiguities are found in applied research. Mansfield and Snyder’s (1995) work on democratization and war, for example, repeatedly invokes conflicts in the former Soviet Union and the Balkans to illustrate their argument. However, conflicts such as Chechnya and Bosnia are not considered international wars and not included in the empirical data they rely on in their analyses. More generally, as a conflict could include many antagonists, civil and interstate wars need not be mutually exclusive. Many wars that occur fully within states have participation by “external” actors outside a state’s border. The conflict in Kashmir has a transnational dimension even if the Pakistani government is not directly involved at a particular point in time. Similarly, the conflict does not cease to become a civil war if forces associated with the Pakistani government intervene directly in the conflict.

### 3 International dimensions of civil war

Whereas research on interstate war has tended to disregard the transnational aspects of wars between states and non-state actors, most empirical studies of civil war have focused on how economic and social attributes of individual states make states more prone to civil wars and largely neglected the role of actors outside the boundaries of the individual state. This is problematic as the risk of civil war of may be influenced by the participants and processes outside the boundaries of the nation state.

In this section, I first discuss some of the broader empirical evidence suggesting that forms of dependence and interaction between states influence the risk of conflict. In the

---

<sup>1</sup>The First Kashmir war was classified as an extra-systemic war in India in the 1994 release of the COW war data, without any Pakistani participation, but is now considered an interstate war between India and Pakistan. Nagorno-Karabakh was initially classified as an interstate war, but has now been reclassified as a civil war in Azerbaijan.

subsequent section, I will examine in more detail some of the relevant linkages between states and how these may affect the risk of civil war.

A wealth of empirical evidence attests to the importance of diffusion for interstate wars (1991). Spatial proximity increases the opportunity for conflictual and cooperative interactions between states as well as the willingness of leaders to engage in interactions. Most wars are fought between neighboring states. Similarly, trade tends to be high between geographically proximate states.

Whereas the importance of geographical diffusion for interstate war is well established, it has in practice received little attention in research on intrastate conflict. Recent work, however, indicates that spatial patterns of spill over and contagion appear to obtain for civil wars as well. Ward and Gleditsch (2002) examined conflict as an autologistic process, where the likelihood of conflict in one country is held to be conditional on the presence of conflict on other states. Their empirical results – based primarily on civil wars – indicated that the risk of war in an individual country is strongly influenced by the presence of wars in other countries. Whereas most models of conflict are unable to predict conflict with a probability above 0.5, the autologistic regression model of conflict examined in Ward and Gleditsch produced predicted probabilities above 0.5 and successfully postdicts many of the conflicts in a sample of data from 1988. Moreover, their model displayed good performance in an out of sample prediction test. A forecast based on the empirical results for 1988 yielded predicted probabilities that identified about half of the wars in the subsequent decade. Moreover, the model did not predict too many conflicts that failed to occur and missed a reasonably small number of actual conflicts.

The good predictive results for model indicate that a conditional model of conflict appears to pick up structural factors influencing the likelihood of conflict, and do not simply reflect idiosyncratic features of the sample used to estimate the model. However, the autologistic regression model in Ward and Gleditsch was based only on data for a single year (1988) and was almost entirely spatial. The model did not include any internal country specific covariates other than level of democracy, and essentially ignored many other domestic factors commonly thought to influence the likelihood of civil war (see the comprehensive review in Sambanis 2002). The Gleditsch and Ward (2002) moreover did not try to identify what specific linkages may underlie geographical contagion of conflict between states in any detail.

Hence, even though the Ward and Gleditsch (2002) study suggests strong international influences on the likelihood of civil conflict, there is a risk that the empirical results may reflect the effects of spatially clustered country attributes that influence the likelihood of war rather than processes and interaction between countries *per se*. I have shown elsewhere that many of the economic and political country attributes which existing research has related civil war such as GDP per capita and democracy also displays geographical clustering (Gleditsch 2002*a*).

In this paper, I develop an empirical model that examines both the role of a larger number of internal country specific attributes as well as differentiates between the various international factors that may influence the risk of civil war.

## 4 Civil war and international linkages

In this section, I develop hypotheses on various external factors that can increase the likelihood of a state experiencing a conflict within its boundaries. Before looking at the different types of external factors, I first need to identify the relations that tie states together and where we may see such external factors exerting an influence on conflict.

The basic argument advanced in this paper is that the likelihood that a state  $i$  will be involved in a conflict at time  $t$  is very much dependent on the processes taking place in the other states. This dependence stems from interaction between states and differences in their relations. Assuming that all states are connected to one another tends, however, provides little indication of why some regions appear to be more conflict prone than others, and tends to lead to intractable empirical models.

Although a single state  $i$  may interact with all of the remaining  $N - i$  states in the international system, not all of these possible relationships between states are likely to be equally relevant. Since distance is such a powerful modifier of the opportunities and feasibility of interactions, closer states can be assumed to be more relevant. We can get a handle on the most important relationships between states by assuming that most of the dependence between states is of a regional character or determined by geographically proximate actors.

One useful simple spatial dependence structure is to assume a local Markov random field. A Markov chain specifies the probability distribution of some discrete variable  $y_{i,t}$  at time  $t$  as a function of the state of observation  $i$  at previous time periods and a  $J \times J$  matrix of transition probabilities between the various  $J$  possible states that the variable  $y_i$  may acquire. A chain is said to be first-order Markov if the transition probabilities depend only on the state at the preceding time period  $y_{i,t-1}$  and are independent of the state at previous  $T$  time periods  $y_{i,t-2}, y_{i,t-3}, \dots, y_{i,t-T}$  (Harary, Norman and Cartright 1965). A Markov random field can be seen as a spatial analogy to a first order Markov property in time. For a set of  $N$  spatial units,  $Pr(y_i | y_j, j \neq i)$  depends only on  $x_j$  if and only if  $j$  is a neighbor of  $i$  (Ripley 1988). The structure of dependence or influences among units can be modeled through a matrix specification of dependence based on the geographical distance among units, where the entries  $w_{i,j}$  of a  $N \times N$  connectivity matrix  $\mathbf{W}$  acquire non-zero entries if the units  $i$  and  $j$  are connected or geographical “neighbors” (Harary, Norman and Cartright 1965).

### 4.1 Intrastate conflict linkages

The likelihood of civil war is directly increased by the presence of wars in neighboring states. The consequences of conflict in one state can induce spill-over effects and alter the prospects for violent conflict in other states. There are multiple different mechanisms by which this may occur.

Many civil wars become internationalized through direct intervention from neighboring states. Third parties may intervene directly in civil wars. Gartzke and Gleditsch (2001) suggest that third parties may intervene to *bandwagon* or shorten an ongoing conflict through increasing the likelihood of victory or settlement, or to they may intervene to *balance* to promote outcomes or settlement that are relatively more favorable to one of the parties.

States may also intervene in indirect ways by providing support to one of the parties in a conflict that may escalate to violence. The presence of conflict in one state may also decrease the relative costs for insurgents in other states, as arms and sources of material support become more available at a lower price (Collier and Hoeffler 1998).

It is difficult to discriminate between the different causal linkages that may underlie cross border contagion of conflict without more disaggregated data. If such linkages apply, however, we would expect to observe a higher likelihood of conflict in state if neighboring states are involved in either civil or interstate wars.

## 4.2 Intrastate political linkages

Although leaders may have many incentives to intervene directly or indirectly in ongoing conflicts in neighboring states they may face potential constraints on their opportunities to do so from other institutions. Research on the democratic peace suggest that leaders in political institutions with a high degree of popular participation may face greater difficulties in intervening if involvement is opposed by other political actors (see, for example, Gleditsch 2002*a*; Tures 2001; Schultz 1998).

The political context prevailing in a region contains information both about the incentives for violent conflict in adjacent states as well as the prospects for leaders to become involved. Democracy is often defined in terms of institutions that have potential power to constrain executives. The more constrained political leaders in a region, then stronger the expected barriers against direct involvement in civil wars in neighboring states (Gleditsch 2002*a*). Accordingly, we would expect that more democratic regions surrounding a state should decrease the risk that conflicts will escalate to violence.

## 4.3 Intrastate ethnic linkages

Many civil wars involve ethnic groups who try to attain autonomy or secede from existing states. External intervention in conflicts are often motivated by states seeking to support members of similar ethnic groups in adjacent states. Similar, ethnic kin and diasporas in other states have often played an important role in financing insurgencies. Collier and Hoeffler (1998) demonstrate a positive relationships between the size of diasporas and the risk of civil conflict. All else equal, we would expect that the risk of civil wars would be higher in a situation where many of the same ethnic groups are found on both sides of international borders.

## 4.4 Intrastate economic linkages

A wealth of recent research suggest that higher economic interdependence between states decreased the likelihood of interstate war. Greater levels of interdependence between states may lead to greater costs of conflict, since conflicts would disrupt economic relations between states (Russett and Oneal 2001). Alternatively, interdependence may provide states with avenues to substitute military forms of conflict with non-military types (Gartzke, Li and Boehmer 2001).

Interdependence can similarly have a limiting effect on conflict between states. Actors in more integrated and complex economies have greater interests in maintaining peaceful

relations and face greater costs in resorting to conflict. Economic interdependence may exert a conflict dampening role even when potential rebel groups are only marginally integrated in the formal economy or distribution of labor. In a situation where levels of interdependence are high and conflict would be disruptive to many actors, these affected interest will have an incentive to lobby governments to find solutions to accommodate agrieved groups. Such pressure for finding settlements for non-violent conflict may occur both within states or between states.

The relevant actors that can exert influence in a particular situation are likely to be regionally confined. Although trade is not a perfect indicator of integration between states, it has the advantage that data are relatively easy to obtain. On the basis of these arguments I hypothesize that the more integration among actors in a region, the more avenues for mediation and settling conflicts in non-violent ways. In Gleditsch (2002a), I found that greater levels of trade integration between states in a region decreased the likelihood of civil conflict.

#### 4.5 Central domestic factors in civil wars

To evaluate whether international influences exert some effect on the likelihood that a country will experience a civil war requires a reasonable baseline model of conflict taking into account domestic factors commonly thought to be associated with conflict.

Many researchers claim that level of development alters the prospects for civil war in an individual states (1998). Greater levels of material wealth may reduce the intensity of conflict between groups. Wealthier societies will tend to have more capable states that are better able to find political solutions to address grievances which may lead to conflict, or alternatively, have more means to effectively repress domestic dissent.

A large body of research has hypothesized that type of political system will exert an effect on the risk of civil war. Civil wars should be relatively less likely in democracies, as these provide greater opportunities for groups to pursue their objectives by peaceful means. However, the prospects for civil war are not necessarily high in strict autocracies. Despite a close political system, many autocratic states are sufficiently repressive that they can quite effectively deter political dissent. Many scholars have argued that the risk of civil war should be highest in anocracies that combine features of both autocracy and democracy. These states have a combination of sufficient grievances and opportunities that make resort to violence feasible and more attractive. The political system is sufficiently closed that groups may be unable to exert influence through other political means, yet not repressive enough successfully deter conflict. This is sometimes referred to as the inverted U-curve hypothesis.

Many researchers have hypothesized that changes in political systems may be more important than the current level of democracy. Mansfield and Snyder (1995) have hypothesized that democratizing regimes may be more prone to conflict than stable autocracies or democracies. (Gleditsch 2002a) found that although democratization tended to reduce the likelihood of conflict between states, democratization appeared to have an ambiguous effect on the likelihood of civil wars. Although higher levels of democracy reduced the risk of conflict, changes towards less autocracy could increase the risk of civil wars, in particular for countries located in regions with many autocracies.

## 5 Data and empirical operationalization

In this section, I detail the variables and measures used to test the hypotheses set forward in the previous sections

### 5.1 Conflict and war data

By far the most frequently used data on civil war are the data compiled by the Correlates of War project. The Gleditsch and War (2002) paper relied on the COW war data, with updates from the conflict data compiled at the University of Uppsala (Wallensteen and Sollenberg 1999), which is available for the 1990s and late 1980s.

The Correlates of War project requires that a conflict must involve at least 1000 battle deaths in a given calendar year to be counted as a civil war. This is a relatively high threshold, which may exclude many major civil conflicts. The dates for conflicts can become somewhat arbitrary, as wars with lower intensity may drop in and out of the sample depending on whether they claim one thousand casualties in any given year.

A new version of the Uppsala conflict data has coded a larger set of conflicts back to 1945 (Gleditsch, Strand, Sollenberg and Wallensteen 2001). In addition to wars with at least 1000 casualties, the new data includes minor armed conflicts with more than 25 deaths in a year as well as an intermediate conflict category. For this study, I will rely on the conflict indicator in these data. The main dependent variable will be a dichotomous indicator of whether a state experiences a conflict in any given year. I will also use a dichotomous conflict indicator restricted to conflicts that reach 1000 battle deaths. The first variable will be denoted  $y$  and the latter will be indicated with the superscript  $y^+$ .

For the purposes of this paper conflict is not necessarily relevant depending on whether it is classified as intrastate or interstate. In the empirical analysis below, I will use both a composite conflict variable, including both interstate and "civil" conflicts, denoted  $y_{i,t}$ , as well as more restricted civil conflict variable  $y_{i,t}^{cw}$  not including the observations defined as interstate by war by the Uppsala data project. Since there are relatively few intrastate conflicts, the civil wars vastly dominate in the composite variable  $y_{i,t}$ . It turns out to make little difference for the empirical results whether the full composite variable or the more restrictive civil war variable.

### 5.2 Country specific covariates

I include a measure of a state's real GDP per capita, measured in constant 1985 dollars, based on the Penn World Tables (PWT). Since the PWT does not contain data for many developing and socialist states, these figures have been expanded by figures based on other available sources such as the CIA world factbook (Gleditsch 2002b). The effect of GDP per capita on conflict is unlikely to be fully linear but matter more when states are relatively poor. I thus use the natural logarithm of real GDP per capita in 1985 prices. This variable will be denoted  $g^r$ .

The type of political system is here operationalized using the composite Polity democracy, which ranges from -10 to 10. This variable will be denoted  $d$ . Values closer to -10 indicate more autocratic polities, whereas a score closer to 10 indicates more democratic polities. Gurr and Jagers (1995) suggest a tri-partite typology of democracies (a score of

six or above on the composite scale), anocracies (between -5 and 5), and autocracies (minus six or below), which can be used to test the U-curve hypothesis.

Political change is here operationalized as changes on the composite Polity democracy over the prior 10 year period. Positive values indicate net changes toward greater democracy and less autocracy whereas negative values indicate change towards more autocratic political institutions. This variable will be denoted  $\Delta d$ .

Much of the literature on conflict has argued that ethnic heterogeneity increases the risk of war within societies. There is little consensus on what aspects of heterogeneity would matter most. Whereas some focus on fragmentation between many groups, other argues that the likelihood of conflict is greater when a dominant majority suppresses minorities. I address the potential for conflict in ethnic heterogeneity by a variable indicating the size of the largest minority based on Vanhanen (2001). This will be denoted  $e_{i,t}$ .

### 5.3 Regional covariates

I measure regional linkages between states by a new data set indicating the minimum distances between the outer boundaries of states (Gleditsch and Ward 2001). I use a threshold of 950 km to determine whether states are connected to one another. Given the linkages between states in the connectivity matrix  $\mathbf{W}$ , we can define a variety of regional covariates reflecting the factors hypothesized to increase the likelihood of conflict in a given state.

The first set of factors are conflict in other proximate states. Given the distribution of civil wars  $y_t^{cw}$  we can define an indicator of presence of a civil war in a connected states as  $r_{i,t}^{cw} = (w_{(i,\cdot)} y_t^{cw}) \#$ , where  $\#$  indicates the Boolean product. This variable  $cw_i$  will acquire a value of 1 if one (or more) of the neighbors  $j$  of  $i$  are involved in a civil war at time  $t$ . Similarly, we can define an indicator of presence of an interstate war in a connected states  $r_{i,t}^{iw} = (w_{(i,\cdot)} y_t^{iw}) \#$ , which will acquire a value of 1 if one of the neighbors  $j$  of  $i$  are involved in an interstate war at time  $t$ .

The second set of regional factors hypothesized to influence the likelihood of conflict pertains to the regime types of other connected state. We can define a variable indicating average level of democracy among states in a region surrounding a country  $i$  at time  $t$  as  $d_{i,t}^r = w_{i,\cdot}^s d_t$ , where the superscript  $i$  indicates a row-standardized connectivity matrix  $\mathbf{W}^s$  where all the entries in each row sum to 1.

The third set of regional factors hypothesized to affect the likelihood of civil war are levels of economic integration. This is defined as  $i_{i,t} = \frac{W_{i,t} T_{i,j,t}}{g^{cp}} \forall j = \{1, \dots, N\}$  or the sum of country  $i$ 's trade  $T$  with all adjacent countries  $J$  as defined by  $\mathbf{W}$ , relative to a country's GDP  $g^{cp}$ . Since trade figures are indicated in current prices, I use GDP in current prices for the trade ratios, superscripted  $cp$ .

The fourth regional factor hypothesized to affect the risk of civil conflict is the number of ethnic groups across borders. I operationalize this using data from the Minorities at Risk project. More specifically, I rely on an indicator of the number of groups in a state that also exist in adjacent countries. This will be denoted  $e^r$ .

### 5.4 Dependence in time

The pooled structure of the data in time raises additional questions about the independence of the observations. Ward and Gleditsch (2002) estimated an autologistic regression on

conflict data for a single year to avoid the added complications of dependence in time. Since conflict is a rare event, a sample based on a single year is unlikely to include many conflicts, and samples from different years may differ considerably. More information is generally better than less, and it would seem advantageous to include observations from as many years as possible.

Many have pointed out the risk of conflict in an individual country may depend upon its prior history of conflict. The risk of recurrent civil war is high immediately after previous conflict, but the stability of peace between parties generally increases with additional years of peace between the parties.

Beck et al. (1998) and Raknerud and Hegre (1997) have suggested that the influence of a country's prior conflict history could be taken into account by including a count of the years a country has remained at peace  $py$ , either since its last conflict or since first being observed in the data. This is often referred to as a "peaceyears" variable. Since additional years are unlikely to contribute much to the stability of peace in countries that have remained at peace for an extended period of time, Beck et al. suggested modeling the impact of time through a non-parametric approach. Raknerud and Hegre suggested an exponential function where a country's time at peace decays relative to a half time parameter  $\alpha$ , i.e.,  $e^{[-py/\alpha]}$ . For models of conflict with cross-national data, the two approaches in practice tend to yield substantively similar. In this paper I include such a function of  $py$  as a covariate in the model to control for the effects of time dependence, primarily because it is easier to interpret than a non-parametric approach. Trial and error suggest that  $\alpha = 4$  provided a reasonable fit to these data. This implies that the risk of recurrent conflict is halved about every five years.

Another alternative is to estimate a transition model indicating how covariates influence transitions in and out of conflict (Beck, Epstein, Jackman and O'Halloran 2001). This is attractive on theoretical grounds, as the effects on conflict settlement may not be mirror images of those on conflict onset. Due to estimation problems I have not been able to consider this approach in the present version of the paper.

## 5.5 Domestic and regional factors in civil war

The model to be estimated hypothesizes that the presence of conflict for a given state  $i$  will be a function of the set of country specific attributes and the regional attributes discussed above.

$$y_{i,t} = f\left(e^{(-py/\alpha)}, d_{i,t}, \Delta d_{i,t}, d_{i,t}^r, g_{i,t}^r, i_{i,t}, r_{i,t}^{iw}, r_{i,t}^{cw}\right),$$

all the variables as defined above. The model can be assumed to be linear in the log-odds

$$z_{i,t} = \alpha + \beta_1 e^{(-py/\alpha)} + \beta_2 d_{i,t} + \beta_3 \Delta d_{i,t} + \beta_4 d_{i,t}^r + \beta_5 g_{i,t}^r + \beta_6 i_{i,t} + \gamma_1 r_{i,t}^{iw} + \gamma_2 r_{i,t}^{cw}.$$

This model is similar a the standard Logit model, but conflict appears on both sides of the model in the sense that conflicts in adjacent states influence conflict in  $i$  through the  $\gamma$  parameters. The individual observations are treated as conditional on one another rather than independent. Given  $z_{i,t}$  we can write the probability of conflict in a country  $i$  as

$$Pr(y_i = 1 | r_{i,t}^{iw}, r_{i,t}^{cw}) = \frac{e^{z_{i,t}}}{1 + e^{z_{i,t}}}.$$

## 5.6 Estimation methods

The model for  $Pr(y_i = 1 | r_{i,t}^{iw}; r_{i,t}^{cw})$  outlined above has a complicated likelihood function since the conflict observations  $y_i$  are conditionally dependent the value of  $y$  in other states. Many techniques have been developed for estimating such conditional models with continuous dependent variables (Anselin 1988). However, the measure of conflict here is a discrete variable. Spatial statistical methods are much less developed for categorical variables.

One alternative to address the intractable likelihood function is to use Markov Chain Monte Carlo (MCMC) simulation methods to approximate the full likelihood function (Besag 1974; Cressie 1991). Ward and Gleditsch (2002) apply MCMC methods to estimate a conditional model of conflict. Although the final goal is to estimate the above model approximating the likelihood function with MCMC, these methods are very computationally intensive, and it has not been possible to complete this for the present version of the paper.

A less computationally demanding alternative to full MCMC approximation is to estimate the model using maximum pseudolikelihood (MPL). This method is easy to implement and has been shown to have reasonable asymptotic properties. The major disadvantage of the pseudolikelihood approach is that it tends to be inefficient, *especially* when spatial interaction is strong (Huffer and Wu 1998). Nonetheless, the MPL results should provide a reasonable first cut to probe whether these external and regional covariates influence the risk of civil war, even when controlling for standard country specific attributes.

## 6 Empirical results

The results of the conditional model estimated using all conflicts irrespective of magnitude are displayed in Table 1. Many of the expectations are strongly borne out by the empirical results. Presence of either civil or interstate war in an adjacent country increases substantially the risk of conflict. This provides strong evidence for spatial clustering in conflict, and is consistent with what we would expect to see if conflict in one state has spill-over effects for other adjacent states.

The other hypothesized regional influences are consistent with the expectations. More democratic regions are less likely to experience conflict. Countries with higher interregional trade are also significantly less likely to experience civil wars. A higher number of ethnic groups that exist across state boundaries increase the risk of conflict.

The effects of the country specific variables are somewhat less consistent with expectations. Larger second minorities increase the likelihood of civil war. Larger GDP per capita decreases the risk of civil war, but the coefficient estimate is not statistically significant. More surprising is the apparent positive effect of democracy. The alternative specification with the tri-partite typology furthermore yielded no evidence for the hypothesized u-curve. The coefficient estimate for democracy was higher than that for anocracies and both were positive. Democratization has a negative coefficient estimate, but this is not statistically significant. Note that the coefficient estimate for regional democracy is of greater magnitude than that of country specific democracy.

Figure 1 displays the marginal effects of changes in the covariates holding other variables at their mean or median as appropriate, and assuming a case without conflict in adjacent

Table 1: Logit estimates for all conflicts

Covariate	Coefficient estimate	Standard error
(Intercept)	-3.8273	0.5349
Conflict history	4.6734	0.1582
Transborder groups	0.0284	0.0145
Democracy	0.0152	0.0094
Regional democracy	-0.0229	0.0160
Change in democracy	-0.0019	0.0106
Share of largest minority	0.0070	0.0036
ln GDP per capita	-0.0354	0.0655
Regional trade	-2.3066	0.7997
Adjacent civil war	0.2645	0.0719
Adjacent interstate war	0.3825	0.0574

LR- $\chi^2=2508.786$  (DF=10), N=5070

countries. (In a case with conflict, the surface of predicted probabilities would be the same shape, but much higher absolute predicted probabilities). As can be seen, the effects of differences on the regional democracy variables seem to be of a greater magnitude than the differences following changes in country specific differences. Although these results suggest that democracies *ceteris paribus* are more likely to experience civil war, the risk of civil war is nonetheless very small for democracies located in more democratic regional contexts.

Trade integration seems to exert a larger effect on civil wars than do differences in GDP per capita. Thus, the regional effects seem to be of a magnitude at least as large as if not larger than country specific attributes. For the ethnic variables, differences in transborder linkages and heterogeneity seem to be of roughly comparable importance.

Although the marginal impact of each of these variables cannot put a case where all the other variables have values around the mean or median, we would expect many of these variables to covary and go together in real world cases. In a heterogeneous world, undemocratic regions and low integration tend to go together, often in zones of protracted hostilities. The idea of integrated security communities as an avenue to stable peace similarly points to situations where many of these factors go together (for a more extended discussion, see Gleditsch 2002a). When the effects of a large number of variables increasing the likelihood of conflict are added together we see much more dramatic effects, and may push cases over the 0.5 level. The likelihood of civil war in an extremely unfavorable region would be several hundred percent higher than the risk of conflict in very favorable neighborhoods. The differences in the risk of civil war between a country in an “unfortunate” region and a relative benign region cannot be fully accounted for by attributes of the individual country. This provides strong evidence in support for that transnational linkages are important influences on the likelihood of civil wars.

It has often been noted that many empirical models fail to generate predicted probabilities above the 0.5 level (i.e., conflict more likely than peace). This model generates predicted probabilities above this threshold for many cases. Figure 2 displays a density plot of the predicted probabilities from the model given the observed data. This plot suggests

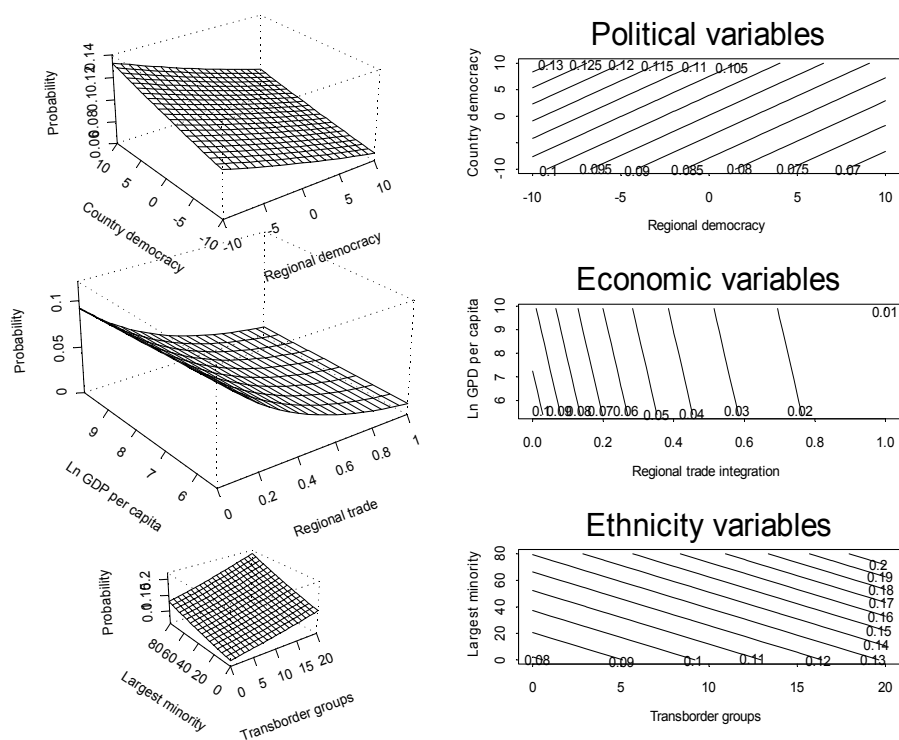


Figure 1: Marginal effect plots

that there are two broad types of groups. By far the highest density is observed around a cluster of observations with low predicted probabilities, indicating that most observations are predicted to be unlikely to experience conflict. However, the distribution also has a second peak, and a substantial number of country years have predicted probabilities of conflict that exceed the 0.5 threshold.

How accurate are the predictions of the model for the observed data? One way to examine this is to look at a simple classification table. This is displayed in Table 5.

Table 2: Classification of conflict states, annual observations

Observed	Predicted	
	No	Yes
No	3624	333
Yes	183	930

As can be seen, the model is quite successful in postdicting conflict and peace for the annual observations. Of all the 1113 conflict years in the model, 930 are successfully predicted to have conflict with a probability greater than 0.5. The model also successfully predicts peace in 3624 of the 3957 annual observations without conflict. In addition to the correctly

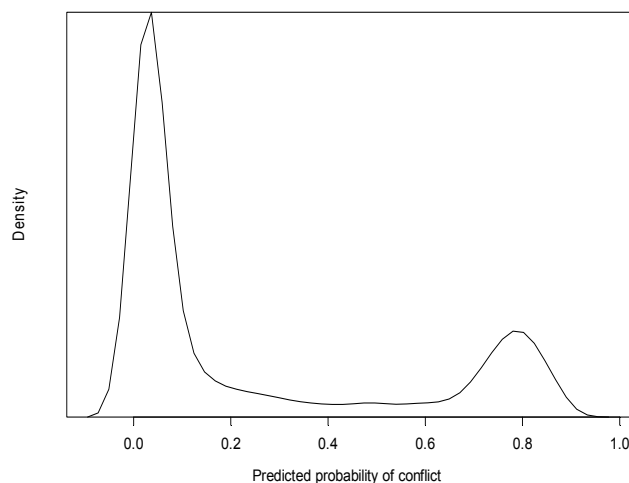


Figure 2: Density Function for Predicted Probability of Conflict. *The density plot of predicted probabilities suggest that there are two broad types of groups. As can be seen, most observations are predicted to have low probabilities of conflict. Nonetheless, a substantial number of country years have predicted probabilities of conflict exceeding 0.5*

classified cases the model also incorrectly predicts conflict in 333 country years and misses 183 actual conflict years. Nonetheless, the model is quite successful in classifying cases of conflict and cases where conflicts do not occur.

Given the relatively low violence threshold in the data, conflict is not such a rare even, and occurs in about 22 percent of all the country years in the sample. An alternative procedure would be to focus only on the major events. The results of estimating the same model restricting the dependent variable to major conflicts that reach the fatality level where the conflict would be considered a war are displayed in Table 3.

The coefficient estimates are generally not dramatically different from those found with the more comprehensive conflict variable. The coefficient estimate for transborder groups now has an incorrect sign. How good of a job does this model do in classifying the major conflict incidents? The classification table in Table 6 suggests that the model does a reasonable job in classifying the major incidents.

I have also experimented with a number of variations on the covariates and operationalizations. The coefficient estimates are generally consistent, but somewhat fragile between different model specifications. This may be due to the inefficiency of the estimation technique. Ward and Gleditsch (2000) demonstrate that the MCMC estimates generally outperform MPL for conflict data where the spatial association is relatively strong. Hence, some caution should be exerted in interpreting the individual coefficient estimates in too much detail. Generally speaking, however, the model appears to pick up on important factors associated with civil wars. The results indicate that the regional context appears to be very important. Many established findings with respect to country specific variables appear not to hold up to the same extent with interdependence within regions and differences between

Table 3: Logit estimates for wars

Covariate	Coefficient estimate	Standard error
(Intercept)	-3.7328	0.5954
Conflict history	4.3796	0.1724
Transborder groups	-0.0105	0.0168
Democracy	-0.0133	0.0109
Regional democracy	-0.0198	0.0184
Change in democracy	0.0232	0.0126
Share of largest minority	0.0069	0.0042
ln GDP per capita	-0.0122	0.0731
Regional trade	-2.0215	1.0061
Adjacent civil war	0.2736	0.0673
Adjacent interstate war	0.5293	0.0879

LR- $\chi^2=1475.216$  (DF=11), N=5070

regions are taken into account.

Table 4: Classification table for war states, annual observations

Observed	Predicted	
	No	Yes
No	4243	256
Yes	150	421

## 7 Forecasting post 1989 conflict

The results in the previous section indicate that the model is quite successful in classifying conflict based on the observed data. This is strictly speaking a “postdiction” rather than a prediction, as it is based on coefficient estimate fitted to the same data which the model is subsequently used to predict. The classification may thus conceivably pick up on idiosyncracies in the estimation sample rather than any causal structure that have some degree of permanence over space and time.

One way to examine whether the model reflects general relationships rather than sample features is to first estimate the model on a smaller subset of the data and then use the coefficient estimates to predict conflict out of sample. To do this, I reestimate the model for country years prior to 1990 and then use these results to predict conflict out of sample for the 1990s.

The coefficient estimates for the sample of pre-1990 years do not differ markedly from those previously reported, and are not shown here to reduce space. Applying these coefficient estimates to the observed post 1990 data yield quite accurate predictions. However, this is not an entirely realistic assessment of the predictive ability of the model in an out

of sample forecast. The problem lies in that allowing the model to use information about the actual occurrence of conflict in all other states  $y_{-i}$  when making predictions about the risk of conflict for individual states  $i$ . In a real forecasting situation, we cannot assume knowledge about conflicts  $y$  in the international system when we make predictions for  $y_{-i}$ .

A better alternative is to use the model and data from 1989 to generate predicted probabilities of conflict that can be tested on post 1990 cases. If we take the predicted probabilities based on these data and use these as predictions about 1990-99, how good will the predictions be? To avoid the added complication of time, I treat the point predictions based on the 1989 data as point predictions about the likelihood of conflict for the 1990-99 period as a whole.

The 1990s has a number of cases where states are participants to conflicts in other states. But if look only at the cases classified as civil wars it is clear that the model is quite successful in identifying where civil wars will occur in the 1990s. As can be seen from Table 5 comparing predictions of the model compared to the actual events in the 1990s, the model correctly predicts 32 of the actual conflicts over the period. The model predicts conflicts in 7 states that failed to occur, and misses 25 cases of conflict.

Table 5: 1990-99 predictions (all levels), based on 1989 data

Civil war	Predicted	
	No	Yes
No	70	7
Yes	25	32

Table 6 shows the predictions restricting the model to only those conflicts that qualify as major wars. As can be seen, a higher threshold of conflict decreases the number of false positives but increases the number of missed conflicts.

Table 6: 1990-99 predictions (major), based on 1989 data

Major civil war	Predicted	
	No	Yes
No	93	3
Yes	24	14

Table 7 displays a list of the civil conflicts predicted by model 1990-99 based on the coefficient estimates and the 1989 data. Many of the cases in the column for the conflicts not predicted by the model appear to be cases where states participate in civil wars not occurring on their core territory. Canada, for example, are held to be involved in a major civil war in 1998 and 1999. Without further information about the conflicts in the Uppsala data and it is difficult to ascertain whether individual conflicts such as this are appropriate or not. The rationale for spatial contagion is primarily dependent upon the location where conflict occurs, not conflict participation. The Kosovo conflict may well have consequences

for Macedonia, but we would not expect Candian participation in peacekeeping operations to increase the likelihood of civil war in the USA.

Most available measures conflict focus on participation in conflict rather than the location where conflict occur. States can choose whether to intervene in conflicts in far away location, but may not be able to insulate themselves from spill over effects from regional conflicts. In this sense, hypotheses about spatial contagion would be easier to evaluate with geogrpahical information about the conflict. Fortunately, a project is currently underway to code geographical information for the incidents in the Uppsala conflict data.

Table 7: Predicted and missed civil wars for 1990-99, using 1989 data

Correct prediction	Missed conflict	False positive
Afghanistan	Algeria	China
Angola	Belgium	Cuba
Bangladesh	Burundi	Vietnam
Burkina Faso	Canada	Mauretania (interstate)
Cambodia	Comoros	Morocco (interstate)
Chad	Republic of Congo	Nicaragua
Colombia	Egypt	Thailand
El Salvador	Guinea-Bissau	
Ethiopia	Guinea	
Guatemala	Haiti	
India	Liberia	
Indonesia	Lesotho	
Iran	Mexico	
Iraq	Mali	
Israel	Nepal	
Laos	Niger	
Lebanon	Papua New Guinea	
Myanmar	Rwanda	
Mozambique	Senegal	
Pakistan	Sierra Leone	
Phillipines	Togo	
Peru	Trinidad	
Russia	United States of America	
Somalia	Venezuela	
South Africa	Yugoslavia	
Spain		
Sri Lanka		
Sudan		
Syria		
Turkey		
Uganda		
United Kingdom		

## 8 Conclusion

Most research on civil wars has focused exclusively on attributes within states, and treated civil wars in one state as independent of conflicts in other states. The shortcomings of the model and data notwithstanding, the relatively high rate of successful to unsuccessful predictions strongly indicates that the regional factors here examined here appear to reflect important aspects of civil war. The processes driving civil wars seem to be strongly influenced by interactions and processes that cross national boundaries. This paper has shown how spatial statistical techniques can be used to capture linkages between states and differences in the regional environment different states face. A more comprehensive approach to civil conflict should consider consider potential causes both in processes within states and interaction between states.

## References

- Anselin, Luc. 1988. *Spatial Econometrics: Methods and Models*. Dordrecht: Kluwer.
- Beck, Nathaniel, David Epstein, Simon Jackman and Sharyn O'Halloran. 2001. "Alternative Models of Dynamics in Binary Time-Series Cross-Section Models: The Example of State Failure." Paper presented to the 2001 Annual Meeting of the Society for Political Methodology, Emory University, July, 2001.
- Beck, Nathaniel, Jonathan Katz and Richard Tucker. 1998. "Taking Time Seriously: Time Series Cross Section Analysis with a Binary Dependent Variable." *American Journal of Political Science* 42:1260–1288.
- Besag, Julian E. 1974. "Spatial Interaction and the Statistical Analysis of Lattice Systems (with Discussion)." *Journal of the Royal Statistical Society, Series B, Methodological* 36:192–225.
- Collier, Paul and Anke Hoeffler. 1998. "On Economic Causes of Civil War." *Oxford Economic Papers* 50:563–573.
- Cressie, Noel A. C. 1991. *Statistics for Spatial Data*. Wiley-Interscience.
- Gartzke, Erik A. and Kristian S. Gleditsch. 2001. "Balancing, Bandwagoning, Bargaining And War: Signaling and Selection Among Third-Party Joiners." Paper Presented at 42nd Annual Convention of the International Studies Association, Chicago, IL, 20-24 February.
- Gartzke, Erik, Quan Li and Charles Boehmer. 2001. "Investing in the Peace: Economic Interdependence and International Conflict." *International Organization* 55(2):391–438.
- Gleditsch, Kristian S. 2002a. *All Politics is Local: The Diffusion of Conflict, Integration, and Democratization*. University of Michigan Press.
- Gleditsch, Kristian S. 2002b. "Expanded Trade and GDP Data, 1946-99." *Journal of Conflict Resolution* 55:in press.
- Gleditsch, Kristian S. and Michael D. Ward. 2001. "Measuring Space: A Minimum Distance Database and Applications to International Studies." *Journal of Peace Research* 38(6):749–768.
- Gleditsch, Nils Petter, Haavard Strand, Margareta Sollenberg and Peter Wallensteen. 2001. "Armed Conflict, 1945-99: A New Dataset." Unpublished manuscript, Peace Research Institute Oslo, October 2001.
- Harary, Frank, Robert Z. Norman and Dorwin Cartright. 1965. *Structural Models: An Introduction to the Theory of Directed Graphs*. New York: Wiley.
- Huffer, Fred W. and Hulin Wu. 1998. "Markov Chain Monte Carlo for Autologistic Regression Models with Application to the Distribution of Plant Species." *Biometrics* 54(2):509.

- Jagers, Keith and Ted R. Gurr. 1995. "Transitions to Democracy: Tracking Democracy's 'Third Wave' with the Polity III data." *Journal of Peace Research* 32:469–82.
- Mansfield, Edward and Jack Snyder. 1995. "Democratization and the Danger of War." *International Security* 20:5–38.
- Pfetsch, Frank R. and Christoph Rohloff. 2000. "KOSIMO: A new Databank on Political Conflict." *Journal of Peace Research* 37:379–391.
- Raknerud, Arvid and Håvard Hegre. 1997. "The Hazard of War: Reassessing the Evidence for the Democratic Peace." *Journal of Peace Research* 34(4):385–404.
- Ripley, Brian D. 1988. *Statistical Inference for Spatial Processes*. Cambridge: Cambridge University Press.
- Russett, Bruce M. and John Oneal. 2001. *Triangulating Peace: Democracy, Interdependence, and International Organizations*. New York: W.W. Norton.
- Sambanis, Nicholas. 2002. "A Review of Recent Advances and Future Directions in the Quantitative Literature on Civil War." *Defense Economics* TBA:in press.
- Sarkees, Meredith. 2000. "Correlates of War Data on War: An Update to 1997." *Conflict Management and Peace Science* 18:123–144.
- Schultz, Kenneth A. 1998. "Domestic Opposition and Signaling in International Crises." *American Political Science Review* 92:829–844.
- Siverson, Randolph M. and Harvey Starr. 1991. *The Diffusion of War: A Study in Opportunity and Willingness*. Ann Arbor, MI: University of Michigan Press.
- Tures, John A. 2001. "Democracies as Intervening States: A Critique of Kegley and Hermann." *Journal of Peace Research* 38:227–236.
- Vanhanen, Tatu. 2001. "Domestic Ethnic Conflict and Ethnic Nepotism: A Comparative Analysis." *Journal of Peace Research* 36:55–73.
- Wallensteen, Peter and Margareta Sollenberg. 1999. "Armed Conflict, 1989-98." *Journal of Peace Research* 36:593–606.
- Ward, Michael D. and Kristian Skrede Gleditsch. 2002. "Location, Location, Location: An MCMC Approach to Modeling the Spatial Context of War and Peace." *Political Analysis* 10(2):in press.