Opening Closure: Intercohesion and Entrepreneurial Dynamics in Business Groups

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The twinned challenge for entrepreneurial groups is recognizing new ideas and implementing them. In one view, connectivity reaching outside the group channels new ideas while closure makes it possible to act on them. We argue that entrepreneurship is not about importing information but about generating new knowledge through recombining resources. In contrast to the brokerage-plus-closure perspective, we identify a distinctive network position, intercohesion, at the overlap of cohesive group structures. The multiple insiders at this intercohesive position participate in dense cohesive ties that provide close familiarity with the operations of the members in their group. But because they are members of multiple cohesive groups, they have familiar access to diverse resources. We first test whether intercohesion contributes to higher group performance. Second, because entrepreneurship is a process of creative disruption, we test intercohesion’s contribution to group instability. Third, moving from dynamic methods to historical network analysis, we demonstrate that coherence is a property of interwoven lineages of cohesion built up through separating and reuniting in an ongoing pattern of interweaving by which business groups manage instability while benefitting from intercohesion. To study the evolution of business groups, our dataset records personnel ties among the largest 1,696 Hungarian enterprises from 1987-2001.

Business groups face two key challenges in their entrepreneurial mode of operation: recognizing sources of novel ideas, and securing the means to implement them. Recent thinking suggests that brokerage ties of connectivity outside the group provide contact with new ideas in the environment; and that cohesive ties of closure within the group provide trust and mutual understanding for implementation (Burt 2005; Uzzi and Spiro 2005; Obstfeld 2005). Whereas this brokerage-plus-closure perspective sees innovation as importing and implementing ideas, we offer an alternative conception of entrepreneurship as recombination. In our view, truly innovative ideas – in the first instance, a fresh conceptualization of the problem itself – are not free-floating outside the group.

Instead of importing ideas or information, the challenge is generating knowledge. It follows that, instead of long distance contact and casual access, the work of recombination requires intense interaction and deeply familiar access to knowledge bases and productive resources. From this perspective, entrepreneurship, as an enabling capacity, proves productive not so much by encouraging the smooth flow of information or the confirmation of fixed identities than by fostering the generative and productive
friction that disrupts the received categories of business as usual and makes possible the
redefinition, redeployment, and recombination of resources. Yet, a simple expansion of
cohesive group membership would not be sufficient: recombination requires interaction
across diversity.

We argue therefore that entrepreneurship in the business group context is driven by the
intersection of cohesive groups where actors have familiar access to diverse resources
available for recombination. In making this argument, we draw on Simmel’s ([1922]
1964) insight that membership in cohesive groups can overlap. With a method that
allows us to identify cohesive – yet non-exclusive – groups, we bring theoretical attention
to the distinctive structural position at the intersection. Our concept of intercohesion is a
theoretical counterpart to Burt’s (1992) concept of “structural hole” but has different
structural properties. Corresponding to our different understanding of the innovative
process and the structural bases for it, intercohesion, as mutual participation in multiple
cohesive groups, provides the requisite familiarity and diversity for access and for action
through a distinctive network topography that is not a summation of brokerage and
closure.

We further argue that entrepreneurship not only has structural properties within a
synchronic dimension but also has dynamic properties along the temporal dimension.
Specifically, entrepreneurial structures are not only creative but are also likely to be
disruptive. Thus, our task is to analyze structural features that predict group performance
and to analyze whether and how these same structural features contribute to or undermine
the continued existence of the group itself.

We propose historical network analysis as a new vantage point from which to re-examine
the essence or constitution of “groupness” itself. The conventional graph snapshot of
network analysis does not distinguish robust and stable collectivities from transitory
alignments; it only enables the distinction of denser or sparser network regions. Within
that conventional framework, it is tempting to focus only on denser regions as cohesive
structures where strong forces of structural determination hold members together. Once
we think of groups as histories of cohesion, however, we can loosen the conceptual hold
of determining structures, and envision groups as sites and tools of agency (Sewell 1992).
We can recognize groups despite temporary losses in density. In fact, we often find
strategic separation between groups within larger units only recognizable from historical
analysis. Our analysis, thus, addresses a fundamental sociological question: What is a
social group across time in network terms?

Our analysis proceeds through three tests. In the first test, we find that intercohesion is a
significant factor explaining outstanding group performance (as measured by high
revenue growth). Moving to dynamic models, in a second test we find that intercohesion
significantly predicts group break-up. Entrepreneurial intercohesion is performance-
enhancing, but comes with the apparent “cost” of group stability. Our findings thus
suggest that intercohesion is doubly generative: its creative tensions of familiarity and
diversity promote group performance in the first instance, and, in the second dynamic instance, these same tensions foster a creative disruption generating dispersion of group members available for later regrouping.

In a third test, we move from dynamic modeling to historical network analysis to analyze the coherence of groups in their repeated reconstitutions. Do de-stabilized groups scatter to the winds or do they reform along patterned lines that would indicate the existence of business groups in the proper sense of historically-constituted larger collectivities? To address this question we develop the concept of lineages of cohesion as a construct that captures the continuous chains of ancestry of the cohesive nuclear structures in our prior models. Our approach furthers an agenda to develop social sequence analysis (Stark and Vedres 2006). Whereas sequence analysis typically conceptualize sequences as isolated and linear, sequences of cohesion are conceptualized here as interconnected and branching. In our case, the “kinship structure” of business group is given by the reproduction and the exchange of members among nuclear groups across “generations” (the years in our study). We find that business groups as larger historical collectivities maintain their coherence through interwoven lineages.

Postsocialist Hungary offers a rich case to examine organizational innovation and the historical evolution of business groups. The dislocations and uncertainties facing state-owned enterprises undergoing privatization, new startups, and foreign-owned subsidiaries can hardly be overstated. With the collapse of COMECON (the Soviet-bloc inter-state trading system) following the demise of the Soviet Union, many firms faced the almost total collapse of their once-guaranteed markets. As the institutional framework to corporate forms was established in 1988, the number of corporations expanded rapidly (from a handful to half a million today). Business networks and business groups emerged quickly as organizational experimentation gathered speed rapidly, to locate, or generate markets for products of the group, and to buffer members from varied uncertainties. Several studies demonstrated the innovative capacity of post-socialist firms, and groups, making use of new corporate institutions, such as cross-ownership, board ties, and holding structures (Stark 1996; Spicer, McDermott, and Kogut, 2000).

In this context, the new institution of corporate supervisory bodies and boards of directors offered firms reliable sources of information, access to insider knowledge of successes and failures elsewhere in coping with challenges, and a mechanism for coordinating actions among strategic business allies. Required by corporate law, the institution was entirely novel to Hungarian executives. CEOs and other senior managers whom we interviewed recalled their puzzlement on attending their first board meeting; but they also emphasized how quickly firms grasped the possibility of sharing directors as an opportunity for coordinating strategy.

Our dataset documents this process of organizational experimentation, the dynamics of group formation and entrepreneurship. We have collected the names of all economic officeholders in Hungary from 1987 to 2001, defining economic officeholders as all
senior managers and members of the boards of directors and supervisory boards of the largest 1,696 companies. With our list of 72,766 names and the precise dates of the tenure of officeholding, we can construct the personnel ties connecting these largest firms for each year in our study. Our case reaches back to the first moments when firms could adopt the newly-legalized, corporate form. It includes periods of business uncertainty involving privatization, transformational recession, marketization, the institutionalization of economic regulations, and massive foreign direct investment (FDI) as well as three parliamentary elections.

**NETWORK STRUCTURES FOR ACCESS AND ACTION**

One prominent feature of business networks is that firms cluster together in cohesive groups. Dense ties among the members of the group, it is argued, provide a basis for trust and a means for coordinating action (Useem 1980; Uzzi 1997). Cohesive ties enable groups to implement projects beyond the capacity of any given firm (Granovetter 2005). The sharing of risk along these ties buffers groups from uncertainty. By mitigating abrupt downturns, risk spreading in such business group networks yields lower volatility of year-to-year profitability (Lincoln, Gerlach, and Ahmadjian 1996).

By another logic, business groups might elect to forego high density within the group in favor of more weak ties to firms outside the group. Such a strategy of sacrificing density for diversity economizes network resources by reducing the number of redundant ties (Burt 1992). Long distance ties provide access to more channels of information outside the group, and this diversity provides a basis for greater adaptability. According to this logic, the conservatizing strategy of in-group cohesion is maladaptive. It risks locking into early successes which will be quickly punished in a situation of rapid change in which the directionality of disruption cannot be foreseen. The defensive strategy of closing ranks risks a false security when the correct choice is not of survival versus innovation, but of innovation in order to survive.

Recent developments in network analysis suggest a third strategy: in place of strengthening ties within the group or reaching outside it, do both. For some researchers in this line of work, the terminology is cohesion and connectivity (Watts 1999; Moody and White 2003; Uzzi and Spiro 2005); others favor the terminology of closure and brokerage (Burt 2005, 2008; Baum, McEvily, and Rowley 2007). Common to them is

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1 Our study contributes to historical network analysis using large datasets across relatively long time frames. Padgett and McLean (2006), for example, develop a multiple network approach to study the genesis of the banking form over some eighty years in Renaissance Florence. Powell, White, Koput, and Owen-Smith (2005) apply a dynamic network approach to understand the emergence of biotechnology, following ties between more than two thousand organizations over a twelve year period. And Uzzi and Spiro (2005) examine network properties of cohesion and connectivity among a population of 2,092 people who worked on 474 Broadway musicals from 1945 to 1989. (See also Wuchty, Jones, and Uzzi. 2007.)
the notion of the complimentarity of these distinctive network properties, especially beneficial where the goal is innovation. Brokerage/connectivity provides access to ideas and information, but in itself lacks the means for implementation. Closure/cohesion provides the means of coordination but lacks diversity for discovery. Together, they can solve the limitations of each, formulated by Obstfeld (2005) as the “action problem” and the “idea problem.” Exemplary of this approach is the recent study by Uzzi and Spiro (2005) who demonstrate that success in Broadway musicals (defining innovation as the ability to produce a “hit”) is a function of enough cohesion (defined as continuity in the composition of the musical “team” from one musical to the next) and connectivity (diversity in composition from one musical to the next).

Our study of business groups in Hungary takes these recent developments as its point of departure. With Burt, Uzzi, and others we share the notion that entrepreneurship (or innovation more generally) is not facilitated by either closure/cohesion or brokerage/connectivity alone. Because we agree that the innovative work of recombination involves a combination of close familiarity and diversity, we similarly seek to identify specific structural features that promote these social processes. However, we depart from the current consensus by arguing for a perspective on entrepreneurial structures that does not involve a summation of the conventional dichotomies. Instead, as we elaborate theoretically in the following section, we develop a conception of intercohesion, a distinctive network structure built from intersecting cohesive groups.

Because we see entrepreneurship as disruptive as well as creative, we are attentive to the finding of Uzzi and Spiro (2005) that innovation requires some reshuffling of groups. Very stable groups (a Broadway musical formed with the same members who worked together in a previous musical) and highly unstable groups (a musical in which few of whose members had worked together before) are less likely to produce an innovation than groups combining members with prior affiliation and novel affiliations. With this insight in mind, we take the next step analytically: is there a structural feature that could predict successful performance and also explain the dynamics of group formation and dissolution?

**INTERCOHESION**

To address this question we must first identify the relevant groups in network terms. Our case differs from those of Burt and Uzzi for whom group boundaries (e.g., the members of a team project, the members of a musical) are given prior to the analysis. Business groups in Hungary – unlike those in East Asian economies where analysts can refer to directories listing the members of Japanese keiretsu or South Korean chaebol – do not exist as named places on the economic landscape.

We adopt a measure of cohesion to identify the components of business groups, with a guiding theoretical principle that cohesive group structures need not necessarily be
exclusive\textsuperscript{2}. That is, we are deliberately attuned to the possibilities that network structures can be cohesive and overlap. As Simmel ([1922] 1964) observed in the \textit{Web of Group Affiliations}, it is frequently the case that a person is a member of more than one cohesive group at a given time. For Simmel, such multiple group membership was a source both of individuation for the person and of social integration for the larger collectivities involved. In our population of firms in Hungary we might find, for example, a power plant linked in a cohesive group with other power plants, while also cohesively tied with power distributors and coal mines, linked in a different cohesive group with other heavy industry companies, and associated in a group with banks.

Recent work in social network methodology acknowledges that actual network structures can be composed of overlapping cohesive groups (Moody and White 2003). Our analytical contribution is to recognize that, if cohesive groups can overlap, there is a distinctive structural position at the intersection. That is, from a methodological residual, we point to the intersecting location as a sociological object worthy of theoretical reflection. We develop a concept, \textit{intercohesion}, to refer to the intersection of mutually interpenetrating, cohesive structures. Actors in this intercohesive position are multiple insiders, participating in dense cohesive ties that provide close familiarity with the operations of the members in their group. But because they are members of more than one cohesive group, they have familiar access to diverse resources. This combination of familiarity and diversity facilitates the work of recombining resources. Intercohesion is closure without being closed off.

Our emphasis on familiarity and diversity does not imply, however, that we see the entrepreneurial work of innovative recombination\textsuperscript{3} as achieved by the summation of closure/cohesion and brokerage/connectivity. Figure 1 illustrates the structural position of intercohesion in contrast to brokerage and closure. Our identification of distinctively

\textsuperscript{2} Clustering algorithms used by network analysts typically parse cohesive structures into separated communities with a resulting blind spot to multiple group membership. In his article on the social group concept, for example, Linton Freeman asserts that overlapping only occurs between groups from different social contexts (such as work, kinship, and friendship), and once the context is clarified, there is little, if any, overlap (Freeman 1992). We argue that partitioning social networks into disjunctive social groups is artificial, driven more by limitations of methodological vision than by sociological insight. Joint appointments in academic departments constitute an overlap of two or more departmental groups. Nuclear families form as the overlap of maternal and paternal kinship groups. It is not exceptional to participate in more than one circle of friends. A more realist perspective thus acknowledges that social groups can be cohesive and overlapping.

\textsuperscript{3} We are drawing on Schumpeter’s definition of entrepreneurship as “the carrying out of new combinations.” “As a rule, the new combinations must draw the necessary means of production from some old combinations … development consists primarily in employing existing resources in a different way, in doing new things with them” (Schumpeter 1934 pp. xxx-xxx). For a neo-Schumpeterian statement in the field of economic growth models, see Weitzman (1998).
intercohesive processes (as opposed to an additive operation of the structure of internal group ties and the structure of external group ties), rests on conceptual differences about entrepreneurship. For the adherents of the additive school, innovation is basically conceptualized as a process of germination: To the nurturing soil of trusting relationships of cohesion, bridging or brokerage ties bring the seeds of ideas and information. We have no doubts that ideas and information are often vital; our question is whether the activity of productive recombination involves only (or even primarily) resources that can flow or circulate.

Figure 1. Brokerage and entrepreneurship.

![Figure 1](image)

We conceptualize entrepreneurial recombination as more complex than the importation and implementation of novel ideas. In our view, the idea problem is itself an action problem. That is, the most innovative ideas are not “out there” in the environment of the group. Instead of waiting to be found, they must be generated (Kogut and Zander 1992). It is one thing to recognize an already-identified pattern; but quite another to make a new association. In this sense, the process of innovation is paradoxical for it involves a curious cognitive function of re-cognizing what is not yet formulated as a category. As John Dewey ([1938] 1998) and the pragmatists argued, it is only in the process of attempting to make a transformation in the world that new problems can even be formulated. Generating novel recombinations is itself a kind of production requiring coordination and cooperation across different communities.

In their study of new product development in cellular telephones, blue jeans, and medical devices, Lester and Piore (2004) demonstrate that each of their cases of radical innovation involves combinations across disparate fields: Fashion jeans are the marriage of traditional workmen’s clothing and laundry technology borrowed from hospitals and hotels. Medical devices draw on both basic life sciences and clinical practice. And cellular phones recombine in novel form radio and telephone technologies. They
conclude that “without integration across the borders separating these different fields, there would have been no new products at all” (Lester and Piore 2004: 14-15).

For us, the telling phrase in this passage is “integration across the borders…” Lester and Piore do not refer to “contacts” across borders, for it is not enough for different communities to be in contact. Recombinant innovation requires that they interact. For these reasons, the conventional pairing of access to diversity is insufficient. Deep access for generating new problems, new knowledge, and new capabilities (as opposed to transferring already accepted ideas) requires considerable trust, hence familiarity. Such access can only be achieved by being an insider, an accepted member of a group. We argue therefore that productive recombination in the business group context requires familiar access to resources that are not provided by the narrow bandwidth of the slender ties of bridging and brokerage.

Yet these ties cannot be so cohesive as to constitute a single group: to be able to recognize the potential for novel recombinations, entrepreneurship requires access to diverse sets of resources – access that is only possible by being a member of two or more cohesive groups. The distinctive structural properties of the intercohesive position provides mechanisms for achieving the diverse familiarities required for re-cognizing resources and for their productive recombination.

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4 Because innovation, in this view, involves bringing together incompatible traditions, we should not expect that the process will be harmonious. With hindsight, it is easy to see that faded fashion blue jeans are a recombination of workmen’s clothing and laundry technology. If we can say that “of course!” cellular phones are the marriage of the radio and the telephone, it is only because, as Lester and Piore show, the respective communities worked from the starting point of their differences.

5 Bridging can only have as much impact on cohesive groups as a two-step path length. Members of bridged groups are at best friends of friends. The strength of impact between groups is limited by the weak ties that are between them. Whereas actors at the structural hole occupy a brokerage position at the gap and tax flows, intercohesive actors occupy an entrepreneurial position at the overlap and recombine resources.

6 In examining how different domain of knowledge are brought together to form something new and original, Lester and Piore argue that “ambiguity is the critical resource out of which new ideas emerge. … The cell phone emerged in the space created by the ambiguity about whether the product was a radio or a telephone; by playing with that ambiguity, the device became something that was different from either of them” (Lester and Piore 2004:54). Lester and Piore further observe that radio and telephone technologies each claim a distinct commercial and engineering tradition, with the segment of the radio industry from which cellular technology was derived being particularly distinctive, based on two-way radio mounted in police cars and fire engines. “The cultural differences between radio and telephone engineering were deep rooted” (Lester and Piore 2004:17).
To test these ideas, we correlate the performance of groups, and the extent to which they are intercohesive. We expect that more intercohesive groups will perform better than more insular ones.

**INSTABILITY AND COHERENCE**

Establishing correlations between structures and performance in a cross-sectional context leaves open the question of durability whether as an transitory moment or a sustainable, self-generating pattern. While testing whether intercohesive structures are performance enhancing, we further analyze whether these same structures are self-perpetuating, self-destructive, exploiting and exhausting group cohesion, leading to lock-in, or promoting structural experimentation. From a theoretical perspective, intersection points are locations of structural tension where multiple routines of operation and schemas to organize resources are at work. As prominent locations of restructuring agency, such intersecting social structures can be engines of social change from within (Sewell 1992).

Our examination of the role of intercohesion in the dynamics of group evolution, moreover, is not unrelated to our conception of entrepreneurship. As Schumpeter observed, alongside innovative recombinations, enterpreneurship contributes to “creative destruction” (Schumpeter 2003). Expressed in network analytic terms, entrepreneurial structures are likely to destabilize groups. As Uzzi and Spiro’s (2005) findings suggest, stability in itself is not the most favorable outcome; indeed, some disruption can be beneficial. We therefore test not only whether the creative tensions of intercohesive familiarity and diversity are performance enhancing in the first instance but also whether these same tensions foster a creative disruption dispersing group members available for later regrouping. We expect that the recombinant opportunities provided by the

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7 Founding statements in the field of network analysis were attentive to issues of duration and temporality. Moreno and Jennings (1937:371), for example, defined cohesion as “the forces holding the individual within the groupings in which they are.” At mid-century, Festinger, Schachter, and Back’s (1950:164) study of social pressures in informal groups addressed “the total field of forces which act on members to remain in the group” (see McPherson and Smith-Lovin 2002 and Friedkin 2004 for discussion). In this emphasis they echoed Simmel, whose publication in an early issue of the *American Journal of Sociology* was titled, “The Persistence of Social Groups” (Simmel 1898).

8 By analyzing the dynamics of intersecting business groups, we follow theoretical insights on structure and agency most prominently formulated by William Sewell (1992). Sociological arguments involving structures are typically aimed at explaining stability, closure, and reproduction. To explain social change, the opening of new possibilities, and ruptures in social reproduction, sociologists look for evidence of external shocks, forces from the outside. Sewell argues for a conception of structure that enables change from within. He argues that the intersection of structures, social groups in our case, helps explain change without reference to external impacts.
intercohesive location will have performance enhancing effects at the group level. We further expect that these benefits come at the cost of group stability.

Overlapping membership can be disruptive of group coordination, relations of reciprocal trust, and a sense of fairness. Those with multiply cohesive attachments might seem to follow strategies that are not transparent to those with single membership, leading to coordination difficulties. In the face of the ambiguous loyalties of the intercohesive actors, other members might suspect they are being exploited or manipulated; and lack of commitment and time devoted to a given group by multiple members might lead to group level dysfunction, perhaps even fracture and fragmentation. Intercohesion, we therefore expect, will be negatively correlated with cohesive group stability.

If intercohesion represents a process of creative destruction, do business groups face a forced choice to aim either at stability or high performance? An historical perspective suggests an alternative whereby business groups keep fragmentation within manageable limits. Viewed in a dynamic year to year time frame, we expect to find a system of membership turnover that appears in considerable flux. But the very notion of business group (and the literature on long-lasting business groups) suggests some degree of coherence lasting more than simply from one year to the next (Mizruchi and Stearns 1988; Granovetter 2005). That is, can considerable instability be squared with the concept of business group as a coherent entity? To address this question, we compliment our dynamic (year to year) analysis with an historical analysis.

In our view the concept of business group is not simply a network analytic construct but an historical network construct. In that perspective, cohesive group structures in a synchronic frame are the building blocks of the larger collectivity constituting the business group in a longer historical time frame. These nuclear groups can be connected across time to other groups through member exchange. By tracing these ties we can construct an historical sociological object shaped by patterns of common ancestry. That is, from nuclear groups we turn to the broader “kinship structures” of business organization.

To identify the patterns of these kinship structures, we develop a concept of lineages of cohesion through which we trace chains of ancestral continuity reaching back in time as much as twelve years. Our intuition is that nuclear groups sharing an ancestry might stay close to one another, connected by inter-cohesive sharing of members and a common, repeatedly interwoven, line of descent. A common ancestry makes group formation easier because routines of collaboration are already familiar. This familiarity facilitates the emergence of intercohesive positions: groups that share an ancestry might be less likely to have radically different routines and cultures of collaboration, thus reducing the coordination costs attendant on the intercohesive position.

We find that some business group lineages operate through a single line of succession. But just as numerous, we also find more complex lineages, branching in one
“generation,” reconsolidating in the next, and repeating these processes of exploration and reconsolidation over successive years with intercohesion playing a decisive role in this splitting and reweaving. Groups connected by member exchange operate as larger collectivities which absorb and hence limit the scattering of members from de-stabilized groups, thus mitigating the disruptions caused by intercohesion. Instead of becoming dispersed across the whole range of groups in the entire economy, groups break up to rejoin with others near them, regrouping their resources in a fresh, yet familiar, combination. Instability thus becomes member recombination. In short, churn is consistent with patterned coherence as business groups cope with the uncertainties of their environment by keeping group fragmentation within manageable limits.

DATA AND METHODS

Data

The dataset that we have assembled includes the complete histories of personnel ties among the largest enterprises in Hungary spanning the years 1987-2001. We define large firms as those listed in the annual ranking of the top 500 firms (based on revenue) for any year from 1987 to 2001. Our inclusion rule results in a population of 1,696 firms. This population of firms represents more than a third of employment, about a half of Hungarian GDP, and almost all of export revenues (Figyelő 2002).

We define economic officeholders as senior managers and members of the Board of Directors and the Supervisory Boards of these large enterprises. Personnel data on economic officeholders were transcribed directly from the official files of the 20 Courts of Registry where Hungarian firms are obliged to register information about ownership and personnel. These registry files contain the names and home addresses of all managers that have signatory rights (entitled to sign documents that become legally binding on the firm). They also register the members of the Board of Directors and the Supervisory Board. For each firm we have recorded the names of all signatories and board members who held office in the period studied, recording as well the exact entry and exit dates of their office holding. Our dataset on economic officeholders contains 72,766 names.

Beyond economic officeholders, we have also collected the names of all political officeholders in Hungary during this same epoch. For the period from 1990-2001, we define political officeholders as every elected politician from the Prime Minister, to the Members of Parliament, to the mayors of all municipalities, including as well the top three levels in the hierarchy of the national government ministries (encompassing cabinet ministers and their political and administrative deputies). For the period prior to free elections, we define political officeholders as all members of the Politburo and the Central Committee of the Hungarian Socialist Workers Party, as well as government
ministers and their deputies, covering the political elite under state socialism. For all periods we record the party affiliation of each officeholder and the exact dates of tenure in office. Our data set on political officeholders in the entire period from 1987-2001 contains 16,919 names. By merging the lists of economic and political officeholders, for any given enterprise in any given month, we can precisely identify whether that company had an economic officeholder who was also a current or former political officeholder affiliated with a particular political party.

We define a personnel tie between two companies if a manager or a board member of one company sits on the board of another company. We record personnel ties as symmetrical, with a start and end date for each tie. The models in this study use an annual time resolution, with personnel ties recorded at the last day of each year. We define a firm as having a political affiliation when one of its economic officeholders is also a current or former political officeholder, and we record the party affiliation of that political officeholder as the party affiliation of the firm. Political affiliations are thus personnel ties connecting firms and parties.

For each firm we also collected data on its annual revenues, capitalization, employment, industrial classification, privatization history, and types of owners (state, domestic private entity, or foreign owner).

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9 Data on political and government officeholders were collected from the National Bureau of Elections (which holds records on all elected political officeholders) and from the Hungarian News Agency (which maintains records on all government officials entering or exiting office). Whereas the Communist Party’s Central Committee is analogous to the parliament of the subsequent democratic period, the Politburo was akin to the role of the government in the later period. Names of political officeholders in these years from 1987-89 were gathered from a comprehensive CD-rom publication (Nyíró and Szakadát 1993).

10 The motivation to include former as well as current political officeholders comes from our interviews with managers of large firms. As one CEO noted: “In Hungary there is no such thing as an ex-politician. Once a politician, always a politician.”

11 We do not record a personnel tie between two firms when the tie is created by a political officeholder. Two firms might invite the same politician or ministry official to sit on their boards, not because they wish to establish a tie to one another, but because they seek a political and/or government connection. Political affiliations are about personnel ties between parties and firms and not between firms. Including a personnel tie between these firms would introduce noise in the data that would potentially blur the patterns of personnel ties created to foster business collaboration.
Identifying cohesive groups with the Clique Percolation Method

To identify cohesive groups we adopt a method that starts from cohesive localities, recognizes groups independent of the global network environment, and identifies intercohesive positions. We use the clique percolation method (CPM) developed by physicists to uncover the overlapping community structure of complex networks (Palla et al 2005), recently demonstrated as a suitable tool to analyze the evolution of cohesive groups (Palla, Barabasi, and Vicsek 2007).

The clique percolation method starts from a clique of $k$ nodes, a $k$-clique. In social network analysis, cliques were often rejected as a useful metric of cohesion because they can highly overlap with other cliques. For example, a network of ten nodes can have two cliques of nine that have eight nodes at their overlap. The standard way of resolving this problem was to parse cohesion by recording the number of times two nodes co-participated in cliques. Use of this similarity as input for cluster analysis yielded exclusive, non-overlapping cohesive regions in the network (Wasserman and Faust 1994).

Whereas standard modeling partitions cohesion at the global network level by counting similarities at the node level, CPM is grounded in strictly local properties at the clique level. Instead of regarding clique overlap as a problem to overcome, it regards clique overlap as the starting point for identifying cohesion. (See Appendix A for details on this method.)

Although developed by physicists, the method improves on standard approaches in social network analysis and resonates with new departures from the conventional models. Most importantly for our purposes, a group identified by the CPM method can overlap with another group. CPM is thus in line with Everett and Borgatti (1998) who recognized the limitations of forced partitioning in various algorithms and pointed to the utility of clique adjacency as a theoretical solution. By relaxing clique membership in favor of clique adjacency, hence capturing group overlaps, CPM achieves greater sociological realism allowing for closer approximation to the notion of community than does the concept of a sociometric clique. In a social community – unlike in a network clique – everyone is not necessarily connected to everyone else. We use CPM to identify groups in all years in our dataset from 1987 to 2001.

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12 As Everett and Borgatti demonstrate, a network of 21 nodes can have as many as 2,187 cliques. The prevalence of overlapping in this case is clearly a far departure from what we would think of as meaningful group processes.
GROUP DYNAMICS AND PERFORMANCE

If intercohesive recombination contributes to entrepreneurship and innovation at the group level, intercohesive groups should outperform exclusive ones. Does intercohesion contribute to group performance? To answer this question we first need to identify a suitable metric of performance, and then isolate the contribution of intercohesion to performance.

Profitability, although a widely-used indicator of performance, has questionable validity in the postsocialist setting. With high taxes and changing government regulations, profits can easily be manipulated – euphemistically, “optimized” depending on prevailing regulations. Additionally, as most firms were undergoing restructuring programs, profitability was not a practical metric to gauge firm performance. Instead of turning immediate profits, the key to survival was obtaining and keeping markets and thus securing revenues. We choose therefore to focus on revenue dynamics.

As a simplest test, we consider the correlation between intercohesion at year \( t \) and revenue growth from year \( t \) to \( t+1 \). Intercohesion is defined for a given group in a given year, as the number of groups with which a given group overlaps. Revenue growth is the growth rate at group level, measured in real terms (corrected for inflation). This correlation between intercohesion and revenue growth is significant (with \( p=.049 \)), and positive, although weak (\( R=.088 \)). This indicates that groups with more intercohesive ties grew slightly faster.

Is higher revenue growth in groups with intercohesion only due to by the firms at the intersection? If intercohesion operates similarly to brokerage, we expect that firms at the intersection will see a faster increase in their revenues, compared to firms with only one membership. To answer this question we computed the same correlation between intercohesion (now at the firm level, as the number of additional memberships held) and revenue growth. This Pearson correlation is not statistically significant (\( p=.127 \)), indicating that the benefits of intercohesion are realized at the group level, and do not appear distinctly at firms occupying the intercohesive position. The mechanisms of intercohesion seem to differ from gatekeeping and brokerage, where benefits – additional revenue generating possibilities – would accrue at the gatekeeping firm spanning the two groups.

To further test the relationship between growth and intercohesion, and isolate this relationship from other predictors of revenue growth, we construct multivariate models. Our goal is to evaluate whether there is a significant relationship between intercohesion and growth even if we introduce variables of intra-cohesive processes (such as group size, or homophily), extracohesive processes (primarily those of brokerage and bridging), and control variables, such as industry composition, and efficiency.
For each group in our dataset, we have a score of real revenue growth. From this continuous variable we construct two categorical dependent variables capturing the distinctive processes of revenue decline and growth. Groups that are more entrepreneurial should achieve higher performance, but entrepreneurship does not guarantee performance. Thus we also expect that more entrepreneurial groups are not buffered from declining performance.

Revenue decline records whether the revenues of the group declined during the year in question. High revenue growth records whether the group belonged to the most successful 25 percent of groups in the overall sample (revenues for this top quartile corresponded to at least eight percent annual growth, controlling for inflation). We have two reasons for transforming our ratio scale variable into two categorical variables. The technical reason is that the distribution of revenue growth is highly skewed – there are many groups with modest growth, and few with extremely high growth. The second reason is substantive – we believe that the predictors of preventing revenue loss, and the predictors of achieving high performance are distinct.
Table 1. Descriptive statistics.

<table>
<thead>
<tr>
<th>Independent variables</th>
<th>Mean</th>
<th>Sd</th>
<th>Min</th>
<th>Max</th>
</tr>
</thead>
<tbody>
<tr>
<td>Intercohesion</td>
<td>2.707</td>
<td>2.746</td>
<td>.000</td>
<td>19.000</td>
</tr>
<tr>
<td>Group stability from t to t+1</td>
<td>.514</td>
<td>.259</td>
<td>.120</td>
<td>1.000</td>
</tr>
<tr>
<td>Group stability from t-1 to t</td>
<td>.438</td>
<td>.282</td>
<td>.000</td>
<td>1.000</td>
</tr>
<tr>
<td>Revenue growth from t to t+1</td>
<td>1.587</td>
<td>3.945</td>
<td>.170</td>
<td>63.190</td>
</tr>
<tr>
<td>Negative growth from t to t+1</td>
<td>.544</td>
<td>.499</td>
<td>.000</td>
<td>1.000</td>
</tr>
<tr>
<td>Top quartile growth from t to t+1</td>
<td>.249</td>
<td>.428</td>
<td>.000</td>
<td>1.000</td>
</tr>
</tbody>
</table>

*Intra-cohesive processes*

| Group size                                    | 4.661  | 1.168  | 4.000 | 11.000 |
| Capital size of the largest firm             | 9.355  | 1.241  | 1.000 | 10.000 |
| Size difference                               | 1.507  | 1.650  | .000  | 9.000  |
| Financial members                            | .722   | 1.101  | .000  | 8.000  |
| Industry homogeneity                          | .409   | .317   | .000  | 1.000  |

*Extra-cohesive processes*

| Brokerage                                     | 19.242 | 10.870 | .000 | 44.000 |
| Bridging ties                                 | 11.217 | 11.806 | .000 | 72.000 |
| State owned proportion                        | .255   | .267   | .000  | 1.000 |
| Foreign owned proportion                      | .253   | .231   | .000  | 1.000 |
| Politicized proportion                        | .223   | .182   | .000  | .800  |
| Political mix                                 | .260   | .439   | .000  | 1.000 |
| Governing party tie                           | .910   | 1.007  | .000  | 5.000  |

*Controls*

| Year                                          | 96.594 | 2.903  | 89.000 | 101.000 |
| Group age                                     | 2.817  | 1.820  | 1.000  | 11.000 |
| Newly formed group                            | .113   | .317   | .000  | 1.000 |
| Labor efficiency (log)                        | .700   | .530   | -.550 | 2.824  |
| Capital efficiency (log)                      | -2.636 | .652   | -4.513| .447  |
| Sum of revenues (log)                         | 4.067  | .701   | 2.400 | 5.701  |

*Industry*

| Energy                                        | .048   | .213   | .000  | 1.000 |
| Mining                                        | .015   | .121   | .000  | 1.000 |
| Chemical                                      | .151   | .466   | .000  | 4.000 |
| Metallurgy                                    | .061   | .253   | .000  | 2.000 |
| Heavy industry                                | .400   | .836   | .000  | 5.000 |
| Light industry                                | .331   | .680   | .000  | 3.000 |
| Wood and textile                              | .140   | .396   | .000  | 3.000 |
| Food industry                                 | .627   | 1.104  | .000  | 8.000 |
| Construction                                  | .258   | .878   | .000  | 6.000 |
| Wholesale                                     | .421   | .655   | .000  | 4.000 |
| Retail                                        | .367   | .659   | .000  | 4.000 |
| Transport                                     | .102   | .319   | .000  | 2.000 |
| Services                                      | .650   | .869   | .000  | 4.000 |
Our first independent variable is intercohesion: the number of groups with which a given group overlaps. We expect intercohesion to lower the probability of a revenue decline, and to increase the probability of high revenue growth.

Our second set of independent variables represent intra-cohesive processes. The first variable is *group size*, the number of firms in the group. Similarly to Simmel’s sociology of numbers, we expect that larger groups will have different revenue dynamics from smaller ones, as it is less likely that a large group achieves extraordinary growth. The second variable registers processes of homophily based on homogeneity in industry profile. We measure this *industry homogeneity* by the numerical difference between the first and second most numerous industry category’s share in the group. If the group is entirely of one industry, this variable is equal to one. If there are two equally numerous industries, this variable is equal to zero. Three other variables refer to processes of economic power and dominance in stabilizing or de-stabilizing the group. *Size of the largest firm* is measured in deciles of capitalization, ranging from 1 (smallest firms) to 10 (largest firms). The expectation is that powerful economic players can hold a group together (Thye, Yoon, and Lawler 2002). To assess the effects of relative economic dominance we record *size difference* as the size decile-difference between the largest and second largest firms in the group. A larger value indicates a more clearly dominant player in the group in terms of size. Whereas Thye et al (2002) expect that the equality of power fosters group stability, Gould’s (2003) formulation would predict that equality of power would lead to conflict and group breakup. *Financial members* records the number of financial firms that are members of a given group. Following (Mizruchi and Stearns 1988), we expect groups with financial members to have inferior performance.

Our third set of independent variables represent extra-cohesive processes. For each group we record *being brokered* as the number of all the other groups to which the group is connected by an intermediary. We also record *bridging ties*, the number of other groups that are reachable from the group in question with a direct tie. These variables represent brokerage as recently re-formulated by Ronald Burt (Burt 2005). Whereas Granovetter (1973) would expect that groups with bridging ties will have higher performance, we are also attuned to the possibility that being brokered has negative implications on performance (Fernandez-Mateo 2007).

---

13 Homophily has been shown to be an important factor contributing to cohesive affiliation (McPherson, Smith-Lovin, and Cook, 2001). McPherson and Smith-Lovin (2002:13) define homophily as “the positive relationship between similarity (on almost any dimension) and the probability that two people will have a network connection between them.” Although McPherson has not studied business groups, the most pertinent dimension of similarity in this context is homogeneity of industry profile. If homophily is operating among our Hungarian firms, cohesive groups concentrated in the same industry should be more likely to exhibit stability than those of greater industrial heterogeneity.
Among extra-cohesive processes, we also consider the reach of the group to the political field, notable in a case that involves profound economic dislocation in the context of simultaneous political transformation (Stark and Bruszt 1998). Business groups are seen as especially suitable features for political affiliation in such emerging markets (Khanna and Rivkin 2001). We include three variables to tap various aspects of these processes. *Politicized proportion* records the proportion of group member firms with party affiliations through personnel ties. A group exhibits *political mix* when we find both left and right party affiliations. *Governing party ties* records a group with a political affiliation to a currently governing party.

The final variables in this set of extra-cohesive processes involve links to owners outside the group.¹⁴ *State-owned proportion* records the proportion of group members with the state as a significant owner. *Foreign-owned proportion* similarly records the proportion of group members in foreign ownership.

As control variables we include *year* as well as *group age*, defined as the average number of years that pairs of group members have spent in groups together. We include specific *industry* categories, an indicator of whether the group was *newly formed* of firms not belonging to any groups in the previous year, *labor efficiency* (measured as revenues over number of employees), and *capital efficiency* (measured as revenues over capitalization). To correct for skewedness of the distribution on these latter two variables, we take the logarithms.

Both models – the model of declining performance, and the model of high performance – pass the Hosmer-Lemeshow test for fit, and Pregibon’s link test of model specification. Tolerance and variance inflation factors were within conventional bounds for all independent variables, indicating that multicollinearity should not be a concern. To test for sensitivity to defining the cutpoint of high performance at top 25 percent, we ran the high performance model with dependent variables representing the top 30, 20, 15, 10 and 5 percent of revenue growth. In all of these models the sign and significance of the intercohesion coefficient was the same.

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¹⁴ Our data set contains detailed information about firms’ ownership structure. For each firm we can record whether it has sizeable state ownership and sizeable foreign ownership, as well as details in the timing of any changes in such ownership. Our definitions of significant state and foreign ownership follow procedures detailed in Stark and Vedres (2006).
Table 2. Logistic regression prediction of revenue decline or high growth.

<table>
<thead>
<tr>
<th>Independent variables</th>
<th>Declining revenue (yes=1)</th>
<th>Top quartile revenue growth (yes=1)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Intercohesion</td>
<td>-.041</td>
<td>.210***</td>
</tr>
<tr>
<td><em>Intra-cohesive processes</em></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Group size</td>
<td>-.469***</td>
<td>-.144</td>
</tr>
<tr>
<td>Capital size of the largest firm</td>
<td>-.046</td>
<td>-.588***</td>
</tr>
<tr>
<td>Size difference</td>
<td>.061</td>
<td>-.159</td>
</tr>
<tr>
<td>Financial members</td>
<td>.536***</td>
<td>-.319</td>
</tr>
<tr>
<td>Industry homogeneity</td>
<td>.753*</td>
<td>-.948**</td>
</tr>
<tr>
<td><em>Extra-cohesive processes</em></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Brokerage</td>
<td>-.026*</td>
<td>-.006</td>
</tr>
<tr>
<td>Bridging ties</td>
<td>.019</td>
<td>-.040**</td>
</tr>
<tr>
<td>State owned proportion</td>
<td>.385</td>
<td>.425</td>
</tr>
<tr>
<td>Foreign owned proportion</td>
<td>-.430</td>
<td>-.355</td>
</tr>
<tr>
<td>Politicized proportion</td>
<td>.984</td>
<td>-2.904***</td>
</tr>
<tr>
<td>Political mix</td>
<td>-.046</td>
<td>-.507</td>
</tr>
<tr>
<td>Governing party tie</td>
<td>-.109</td>
<td>.376*</td>
</tr>
<tr>
<td><em>Controls</em></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Group stability from t-1</td>
<td>-1.685***</td>
<td>.342</td>
</tr>
<tr>
<td>Year</td>
<td>-.064</td>
<td>-.037</td>
</tr>
<tr>
<td>Group age</td>
<td>.033</td>
<td>.162</td>
</tr>
<tr>
<td>Newly formed group</td>
<td>-.480</td>
<td>-.049</td>
</tr>
<tr>
<td>Labor efficiency (log)</td>
<td>-.980***</td>
<td>1.385***</td>
</tr>
<tr>
<td>Capital efficiency (log)</td>
<td>-.189</td>
<td>.563**</td>
</tr>
<tr>
<td>Sum of revenues (log)</td>
<td>-.446*</td>
<td>-.308</td>
</tr>
</tbody>
</table>
In Table 2 we see that intercohesion does not buffer against revenue decline, but it is a strong predictor of high growth. Recall, by definition, that all the groups that we examine in Table 2 are cohesive. Intercohesive groups outperform their counterparts that lack this ambiguous, yet recombinative advantage.

Turning to intra-cohesive variables, we see that larger group size makes decline less likely without contributing to growth. For business groups there is safety in numbers – smaller groups are more at risk of decline. Having a dominant member of large size makes it more difficult for a group to achieve high growth, probably because for large firms a high rate of growth means a large increase in revenue volume. Homophily is a disadvantage. Industry homogeneity increases the probability of decline and decreases the probability of high performance. Groups of more heterogeneous composition are advantaged. This does not apply, however, to groups including members from finance. Groups with financial members are significantly more likely to face declining revenues.
This result might not be surprising to scholars who study business groups: ties to banks are often associated with financial troubles and decreasing performance.

Looking at personnel ties reaching to other business groups, we see that effects on performance diminish as distance increases. Groups with more two-step ties to other groups (mediated through brokers), are slightly less likely to decline, but without any advantage in high performance. Groups with ties at the closest reach, not even one step away but through the interpenetrating ties of intercohesion, outperform groups that are exclusive regardless of how intensively or extensively they are embedded. The number of bridging ties to members of other groups decreases the probability of high performance. This finding is again in a sharp contrast with our findings on intercohesion. Bridging ties – circuits for the circulation of ideas – do not contribute to group success. This is in sharp contrast to the positive contribution of intercohesion. This finding underscores the importance of entrepreneurial generation of ideas through recombination, as opposed to the reliance on imported ideas from other areas of the network.

Playing party politics is tricky business. While political contacts do not offer protection against decline, only the most narrow, highly targeted, strategy has payoffs. More politicized groups have little chance of achieving high growth. Groups that are overly committed to a party put in jeopardy the trust of their business partners. Groups can only benefit from party politics when they have a tie to a currently governing party.

Intercohesion and group stability operate as mirrored opposites. Group stability buffers groups from revenue decline, and it does not contribute to top quartile revenue growth. Stability in group membership enables reciprocity, solidarity, and mutual assistance to act as a kind of safety net, preventing severe market loss in the group. But trust and improved communication within the group are not such assets that stimulate high levels of growth: stability in itself can be a conservatizing closure.
INTERCOHESION AND GROUP STABILITY

Having analyzed how intercohesion is related to group performance, we turn to questions of group stability. To define and measure group stability, we refer to Simmel’s “Persistence of Social Groups” where he argued that it is meaningful to speak of group identity despite shifting membership and low institutionalization if there is some membership continuity in contiguous stages:

We may express this schematically as follows: If the totality of individuals or other conditions in the life of the group be represented by \( a, b, c, d, e \); in a later moment by \( m, n, o, p, q \); we may nevertheless speak of the persistence of identical selfhood if the development takes the following course: \( a, b, c, d, e - m, b, c, d, e - m, n, c, d, e - m, n, o, d, e - m, n, o, p, e - m, n, o, p, q \). In this case each stage is differentiated from the contiguous stage by only one member, and at each moment it shares the same chief elements with its neighboring moments (Simmel 1898: 670-1).

We draw on this insight in our analysis of group stability. For the first year of our dataset we use CPM to identify each of the cohesive groups of firms (for example, cohesive group 1 composed of firms \( a,b,c,d,e \); cohesive group 2 composed of \( m,n,o,p,q \); and so on). For the second year we identify the cohesive groups existing at that time (following Simmel, call them, \( m,b,c,d,e \); \( a,n,o,p,q \); \( v,w,x,y,z \); and so on.) Because network formation is slow at first and the number of groups appearing at the very beginning is only very small, we modify Simmel’s scheme. Instead of simply following the first established groups, we identify all the groups that exist from 1987-2001. That is, we identify and record, for each year, all of the cohesive groups that exist in that year. By observing the composition of all groups in \( t_1 \) and those in \( t_2 \), we can record the proportion of the members of any given group that remained cohesively tied – which is our metric of stability – for all pairs of years.

To measure the stability of groups we record the flow of members between all groups in adjacent years. A group is completely stable from one year to the next if all members of the group identified in the first year appear together in a group in the next year. At the other extreme, a group dissolves if none of the members at \( t_1 \) appears in any groups at \( t_2 \). Between these extremes, a group at \( t_1 \) can split into various sized pieces that are present in groups at \( t_2 \). To measure such intermediate levels of stability we score the average size of the pieces from \( t_1 \) that appear in groups at \( t_2 \), normalizing for the size of the source group.

To test whether intercohesion has its own predictive power in the context of competing explanations we use a multivariate regression model, with group stability as the dependent variable. The independent variables are the same as those that we used in our models of group performance.
We find a negative correlation between intercohesive ties and group stability. The mean correlation across all years is -0.55, ranging from -0.37 to -0.70. This regression model passed Pregibon’s link test of model specification; tolerance and variance inflation factors were within conventional bounds for all independent variables: multicollinearity should not be of concern. Table 3 presents the results of our regression models predicting group stability.

Table 3. Linear regression prediction of group stability.

<table>
<thead>
<tr>
<th>Independent variables</th>
<th>Group stability</th>
</tr>
</thead>
<tbody>
<tr>
<td>Intercohesion</td>
<td>-.018***</td>
</tr>
<tr>
<td><strong>Intra-cohesive processes</strong></td>
<td></td>
</tr>
<tr>
<td>Group size</td>
<td>-.028**</td>
</tr>
<tr>
<td>Capital size of the largest firm</td>
<td>-.009</td>
</tr>
<tr>
<td>Size difference</td>
<td>.010</td>
</tr>
<tr>
<td>Financial members</td>
<td>.006</td>
</tr>
<tr>
<td>Industry homogeneity</td>
<td>.029</td>
</tr>
<tr>
<td><strong>Extra-cohesive processes</strong></td>
<td></td>
</tr>
<tr>
<td>Brokerage</td>
<td>-.008***</td>
</tr>
<tr>
<td>Bridging ties</td>
<td>.001</td>
</tr>
<tr>
<td>State owned proportion</td>
<td>-.078*</td>
</tr>
<tr>
<td>Foreign owned proportion</td>
<td>.061</td>
</tr>
<tr>
<td>Politicized proportion</td>
<td>.021</td>
</tr>
<tr>
<td>Political mix</td>
<td>.004</td>
</tr>
<tr>
<td>Governing party tie</td>
<td>.003</td>
</tr>
<tr>
<td><strong>Controls</strong></td>
<td></td>
</tr>
<tr>
<td>Group stability from t-1</td>
<td>.204***</td>
</tr>
<tr>
<td>Year</td>
<td>.023***</td>
</tr>
<tr>
<td>Group age</td>
<td>-.013*</td>
</tr>
<tr>
<td>Newly formed group</td>
<td>.127***</td>
</tr>
<tr>
<td>Labor efficiency (log)</td>
<td>.037*</td>
</tr>
<tr>
<td>Capital efficiency (log)</td>
<td>-.033**</td>
</tr>
<tr>
<td>Sum of revenues (log)</td>
<td>.004</td>
</tr>
</tbody>
</table>
Table 3. (Continued)

<table>
<thead>
<tr>
<th>Independent variables</th>
<th>Group stability</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Controls</strong></td>
<td></td>
</tr>
<tr>
<td><strong>Industry</strong></td>
<td></td>
</tr>
<tr>
<td>Energy</td>
<td>-.041</td>
</tr>
<tr>
<td>Mining</td>
<td>.086</td>
</tr>
<tr>
<td>Chemical</td>
<td>.032</td>
</tr>
<tr>
<td>Metallurgy</td>
<td>-.003</td>
</tr>
<tr>
<td>Heavy industry</td>
<td>.029**</td>
</tr>
<tr>
<td>Light industry</td>
<td>.045***</td>
</tr>
<tr>
<td>Wood and textile</td>
<td>.068***</td>
</tr>
<tr>
<td>Food industry</td>
<td>.031**</td>
</tr>
<tr>
<td>Construction</td>
<td>.057***</td>
</tr>
<tr>
<td>Wholesale</td>
<td>.032*</td>
</tr>
<tr>
<td>Retail</td>
<td>.019</td>
</tr>
<tr>
<td>Transport</td>
<td>-.009</td>
</tr>
<tr>
<td>Services</td>
<td>-.001</td>
</tr>
<tr>
<td>Constant</td>
<td>-1.613***</td>
</tr>
<tr>
<td><strong>N</strong></td>
<td>467</td>
</tr>
<tr>
<td><strong>Adj. $R^2$</strong></td>
<td>.472</td>
</tr>
<tr>
<td><strong>F (df)</strong></td>
<td>13.671 (33)</td>
</tr>
<tr>
<td><strong>P-value</strong></td>
<td>.000</td>
</tr>
</tbody>
</table>

*: p<.10; **: p<.05; ***: p<.01

We see that multiple membership ties of intercohesion decrease group stability. This finding suggests that the presence of multiple insiders stresses the fabric of cohesion. Actors who are ambiguously committed produce destabilizing tensions inside these groups. The interpenetrating ties, through which mutually intercohesive groups adhere, disrupt stability. Intercohesive adhesion erodes and disperses group cohesion.

Of intra-cohesive processes only group size is relevant: larger groups are less stable. Of the various extra-cohesive linkages only two are relevant to group stability: brokerage and state ownership. State ownership de-stabilizes groups, which is a manifestation of the privatization process: large numbers of state owned enterprises were reorganized and sold. This reorganization erodes the stability of groups with state owned firms.

The number of brokered ties to other groups is significantly correlated with decreased group cohesion, suggesting that brokers adversely impact the structures they exploit. This finding is in line with the idea that the price of brokerage is borne by those that are connected by the broker (Fernandez-Mateo 2007). Our finding show that in addition to
material losses, another externality of brokerage: the erosion of brokered structures. This structural erosion can eventually diminish the opportunities of brokerage themselves.

Among the control variables, we find, not surprisingly that group stability in the preceding year is related to stability in the current year. Group stability has inertia. We also find a positive trend toward greater group stability leading out of the postsocialist period – groups are more stable in the later years. While newly formed groups, without a pre-history of cohesion are stable in the first year of their existence, there is a slight disadvantage to old groups.

Groups with higher labor efficiency are more stable, while groups with a higher capital efficiency are less stable. A group with high amount of capital and few employees is much more stable than a group with low capitalization and many employees. This latter type of group was the typical target of reorganization, disrupting group continuity. In addition, several industry categories are significant, with varying levels of group stability compared to the reference category of agriculture.

To summarize, we found that intercohesion is disruptive: groups where membership is not exclusive suffer a loss of stability. Interpenetrating contact between business groups is destructive even beyond what we would expect by chance – groups seem to break down more often if one or more of their members takes on multiple affiliations.

**RECOMBINANT LINEAGES OF COHESION**

Intercohesion represents a process of creative disruption. Business groups seem to face a fundamental contradiction, a forced choice to aim either at stability or high performance. Choosing exclusivity can yield stability with attendant protection from failure but not high performance. Choosing intercohesive non-exclusivity, on the other hand, opens entrepreneurial possibilities and the promise of high performance, yet at the risk of disintegration. This leaves groups with two possibilities: they either stay exclusive and stable, content with modest, although secure performance, or they engage in intercohesive linking, possibly outperforming most other groups before they disintegrate.

However, our data indicates a third possibility which renders creative disruption manageable: groups connected by member exchange operate as larger collectivities which absorb and hence limit the scattering of members from de-stabilized groups, thus mitigating the disruptions caused by intercohesion. Instead of becoming dispersed across the whole range of groups in the entire economy, groups break up to rejoin with others near them, regrouping their resources in a fresh, yet familiar, combination. Instability thus becomes member recombination.

Our analyses thus far have worked with data that record changes from one point in time, \( t \), to the next, \( t+1 \). In this section we turn from a dynamic approach to an historical one...
by following the pattern of member flows across the entire epoch. Whereas in preceding sections the unit of analysis was a group at a given point in time and we estimated stability and effects on performance at the $t$ to $t+1$ interval, in this section our unit of analysis is a collectivity of groups linked across time and we follow the entire history of these collectivities through the historical structure of membership flows.

From nuclear groups we now turn to broader kinship structures of business organization. The flows of members trace lineages among cohesive groups. Even though a group might not share any members with another in the past, it can still be connected to it through a chain of ancestry. Our intuition is that groups sharing an ancestry might stay close to one another, connected by intercohesive sharing of members and a common, repeatedly interwoven, line of descent. A common ancestry makes group formation easier because routines of collaboration are already familiar. This familiarity facilitates the emergence of intercohesive positions: groups that share an ancestry might be less likely to have radically different routines and cultures of collaboration, thus reducing the coordination costs attendant on the intercohesive position.

When are two groups related by a link of descent? According to Georg Simmel’s (Simmel 1898) idea of the persistence of identical selfhood in groups, a group of five members, $a,b,c,d,e$, is strongly related to a subsequent group with $m,b,c,d,e$ members. It is also clear that a group with $a,b,c,d,e$ is not related to a subsequent group of $f,g,h,i,j$. In between fully related and unrelated there are degrees in strength of a lineage tie. Being faithful to the founding impetus of social network analysis, we believe that the properties of a group are not a summary of the properties of its individual members. Instead, they emerge from the structure of interactions among members. This fundamental insight informs our definition of a lineage tie. It is doubtful that two groups can be related – in the sense that there is continuity in how the group operates and the purpose it serves – if they share only one member. Because group qualities stem from interaction along the ties that the group contains, it takes at least two members to establish continuity between their old group and their new group. Continuity in codes of communication and collaboration depend on some continuity in interaction. Therefore, continuity in trust and in routines of reciprocation and resource sharing cannot be transmitted by one member alone. In line with this insight, we define a *lineage tie* between a group at time $t$ and another group at time $t+1$ as the sharing of at least two members.

The idea of lineage ties extends the notion of the persistence of groups. Lineage ties link groups at adjacent points in time, as ancestors or descendants. A *lineage of cohesion* represents a separate evolutionary path. Persistence concerns the length of the lineage: a group that persists over a long period, even if members are replaced along the way, is connected back in time to a long chain of groups. Persistence concerns only one dimension of a lineage – its width. Lineages can also have thickness – involving multiple groups at a given point in time, all connected by a shared ancestry. While some lineages are simple chains of linear persistence, other lineages are more complex, with branching and reunification. In this case, the structure of lineage ties is more complexly organized:
members leaving a group will have a strong tendency to reform a group with others from the same lineage.

Figure 2 presents an example of an *interwoven lineage*. In 1995 this lineage starts with one group that has three intercohesive ties to groups outside the lineage. In the following year, this group splits into two groups, each of these having two intercohesive ties to one another, as well as having several overlaps with outside groups. Going forward, the two groups are not stable, but they do not dissolve entirely; the lineage is continued. The groups exchange members, while retaining some, so that in 1997, the following year, there are again two groups but of differing composition from the previous year. These groups are interconnected by overlapping membership with two intercohesive ties between them. In 1998, these two groups split into three groups, linked again by shared members but without intercohesive ties to any outside groups. The three groups then merge into two in 1999, and by 2000 these two have merged once again into one group.

![Figure 2. An example of an interwoven lineage.](image)

*Note:* Circles represent groups, with size proportional to membership. Solid arrows represent member continuity of at least two firms, a dashed line indicates an intercohesive group intersection.

The lineage presented in this example displays some distinctive structural properties. Later generations of groups, for example the two groups in 1999, share their ancestry of cohesion all the way back to 1995 (in the postsocialist context, an entire epoch). Moreover, beyond simply sharing the legacy of group routines from 1995, their lines of descent were repeatedly interwoven in such a way that they can, in fact, both trace a lineage to almost all of the preceding groups. This interwoven lineages contains groups that are linked by intercohesion at each point in time. These groups are not stable: almost
all of them split up from year to the next. Although the membership of individual groups is volatile, the membership of the lineage is stable. This stability is not stasis: at any point in time, groups within the lineage are recombined.

We argue that the existence of interwoven lineages follows from an organizing principle that retains members of de-stabilized groups close to one another. But lineages – straight or interwoven – might also form when ties are created at random, i.e., when a firm in a cohesive group has an equal probability of being a member of any of the groups in the coming year. To test clustering under such random conditions, we compare the observed size distribution of lineages to the size distribution in simulated datasets. If we find that similarly sized lineages emerge often in random data, we can reject our hypothesis that lineages are a result of business organizing.

We ran simulations to test randomness as an explanation for the existence of lineages. In our simulations we fix the degree distribution of incoming and outgoing lineage ties in each year, while randomizing the particular connections. In such a simulated dataset, splitting occurs with the same probability in each year as in the observed dataset in that given year, but members that split apart choose their target groups at random. The same holds for mergers – mergers happen with the same frequency as in the observed dataset, but the particular mergers are randomly generated. Each simulation yields a complete lineage dataset from 1987 to 2001. We generated 1000 such datasets to estimate the distribution of lineage sizes when members migrate to other groups randomly.

Our findings indicate that it is very unlikely that the observed lineages are a result of randomly connected ties. As an illustration, Figure 3 shows the ten largest lineages in a representative simulated dataset (where clustering statistics were closest to the mean of all simulations). Figure 4 shows the ten largest lineages in the observed dataset. The most striking difference between the simulated and the observed lineages is that the largest lineage in the simulated data is much larger than the second largest. Our first measure of lineage size distribution is the relative size of the largest lineage to the second largest. In our simulations the largest lineage on average is 32 times the size of the second largest lineage. In our observed dataset the largest lineage is about twice as large as the second. Only in eight of the one thousand simulations was this ratio smaller than or equal to two.

Another measure of clustering is the relative size of the largest lineage component to the size of the complete dataset. In the observed data the largest component encompassed only 14 percent of the total dataset. In the one thousand simulated datasets the largest component on average occupied 69 percent of a given dataset. Only four of the one thousand simulations resulted in a largest component as small as the observed 14 percent. Lineages are not accidents.
Figure 3. The ten largest simulated lineages of cohesion.

Figure 4. The ten largest observed lineages of cohesion.

The interwoven lineage of “Heavy Metal”
We also found that interwoven lineages have dense intercohesive ties among the groups they contain. Looking at all the pairs of groups where intercohesion is possible (groups that coexisted in the same year), we found that the density of intercohesive ties is much higher for those pairs of groups that are within the same lineage. The density within interwoven lineages is 49 percent, i.e., almost half of the pairs of groups within such a lineage are connected by intercohesion. For example, in an interwoven lineage with four groups, three of the six possible pairs would be connected – enough to make the lineage a connected component. The density for pairs of groups that do not share a lineage is only three percent. As a comparison, we computed the same density indices for the typical simulated lineage dataset displayed in figure 3. In this simulated dataset the intercohesion density within lineages is only five percent – in contrast to 49 percent in the observed data.

In Figure 4 we have also highlighted one of the interwoven lineages. This is the steel industry group “Heavy Metal” where Stark (1996) conducted ethnographic research during the early postsocialist period. This group is an interesting illustration of our social sequence method for identifying business groups across time. Heavy Metal is one of the relatively few business groups that exists as a named entity on the landscape of the Hungarian economy. In fact, the names of each firm indicates their group belonging. Of the 18 firms with the Heavy Metal designator in our dataset, our method classifies 17 of them in this interwoven lineage. The history of this business group that started in 1992 (and was studied by Stark in 1993-94) shows that organizational experimentation was not restricted to the early postsocialist period. Through repeated recombinations of subgroup memberships, the coherence of Heavy Metal is maintained amidst changes in ownership and inflows of foreign direct investment as well as across changes in government and shifting political affiliations.

In the context of business groups, a focus on lineages of cohesion highlights that the organizing principle is more complex than bringing firms together into close cohesive contact. The organization of business groups also involves strategic separation – keeping sets of members apart yet, at the same time, maintaining historical coherence. The historical unfolding of organizing business groups leaves its traces as lineages of cohesion. Lineages highlight an important balancing between two forces. While the friction of intercohesion repeatedly dismantles groups, the shared lineage keeps members within a sphere of exchange in which member resources are redistributed and recombined. Intercohesion prevents groups from freezing into the defensive buffering of exclusivity, while lineage prevents groups from exploding, dissipating the resources they accumulated.
CONCLUSION

Social network analysis has produced a rich array of analytic concepts and powerful methods for studying structural features of economic action. From Granovetter’s (1985) concept of embedded action, for example, sociology developed a systematic approach to predicting economic behavior from the structure of social ties. Methods of social network analysis to identify cliques, cores, clans, and other cohesive groups provided powerful operationalizations to the embeddedness concept. With White’s concept of structural equivalence came the idea that actors might behave similarly not because they are linked together but because they share a common location within the more general pattern of ties. Using the method of block modeling, White and his colleagues turned attention from the conventional understanding of networks as “who knows whom?” to argue that absent ties were of equal importance (White, Boorman, and Breiger 1976). Along this line of thought, network analysis extended attention to patterns in the holes or missing spaces in social structure. Granovetter’s (1973) idea of weak ties and Burt’s (1992) concept of structural holes developed these insights into systematic analyses of brokerage opportunities, access to information, and structural constraint. Recently, Watts’ concept of small worlds highlighted the joint importance of cohesive linking and long distance ties reaching across structural voids, to understand flows of information (Watts 1999) and robustness in response to crisis (Dodds, Watts, and Sabel 2003).

Common to all these endeavors is attention to network topology. Each of the concepts – embedding, structural equivalence, weak ties, structural holes, small worlds – refers to specific topological network features. Our work contributes to these efforts by developing the concept of intercohesion and then identifying its corresponding topological feature where cohesive group structures fold into each other. While retaining the insight that structural properties are built from the presence and the absence of ties (the absence of ties across groups, exclusive of the intercohesive position, is significant in defining the groups as distinct units), we point to a distinctive position where cohesive network structures overlap and interpenetrate. Whereas Burt’s conception of structural hole identifies a network location for brokerage, intercohesion identifies a network location for the entrepreneurial activity of recombination.

Thus, in contrast to Burt’s image of a structure that bridges or spans a hole, we consider a site where structures fold together. Behind this difference of imagery lies a further difference in conceptualizing what is transpiring within networks. In the case of brokerage, social networks are channels, a means of transportation, a system of communication. Like electricity circuits, they conduct: The counterparts of electrons are above all pieces of information, but also rumors and resources, that flow through the network circuitry. In that conception, network positions are important to the extent to which they are irreplaceable in maintaining flows between larger segments of the network.
If the flow or movement of information is the critical activity occurring at the conduit or contact point across the structural hole, it is the generation of knowledge through recombination that is the critical activity occurring at the site of the intercohesive position. Correspondingly, in place of the electricity metaphor, the appropriate metaphors are those from molecular chemistry. Instead of networks as the wiring through which informational electrons flow, think of networks as a kind of molecular bonding in which ties connect nodes into larger groupings that represent a new molecular quality and not merely an extension of circuits to further atoms of the network. Intercohesion establishes strong connections between network molecules to generate a more complex material of creative alliances. In this view of a social network, flows are secondary. Information can be codified; but knowledge is a practice, bound up with particular socio-technical ensembles. And the more innovative the knowledge, the less likely it is to be fully codified and easily transmitted apart from these ensembles. Through the overlapping of strong ties bonding to more than one group, intercohesion provides opportunities for mixing or recombining knowledge practices. Intercohesion represents the location where new ideas are generated, as opposed to a location where informational flows are taxed.

Whereas social network analysis has given us a plentitude of concepts for studying the structural properties of networks, the field has yielded far fewer concepts for analyzing the temporal or historical features of networks. This shortcoming could not be predicted from the field’s foundations, for indeed, Georg Simmel, one of the founding figures of network analysis, addressed the temporal dimension in one of the first issues of the American Journal of Sociology in 1898. In his “Persistence of Social Groups” Simmel pondered whether it was meaningful to speak of group identity in cases of shifting membership and low institutionalization.

Simmel’s challenge persists. More than a century later, social network analysis faces a fundamental theoretical and methodological problem when moving from a cross-sectional to a dynamic and historical concept of group cohesion: How can we identify a group across time in network terms? Can it change its composition and still preserve its group identity? The problem is simple where groups are named, catalogued, and registered, e.g., the Supreme Court, the Youngstown Garden Club, the Mitsubishi keiretsu, or the Samsung Group. Despite membership changes over a century, the Department of Sociology at the University of Chicago remains such because of institutional continuity. But the social networks forming Hungarian corporate groups, like many of the groups studied by network analysts, lack such institutionalization. How do we study group evolution where groups might not be named?

Our work addresses this challenge directly. Drawing on Simmel’s insights, we elaborate a conception of historical group identity based on membership continuity in contiguous stages. Such membership continuity need not be absolute, but there must be some overlap (at least two) of group membership to bequeath groupness from one year to the next. The resulting operation yields another new way of representing network topology – a lineage
of cohesion – in which structure can be displayed and analyzed across the temporal dimension. As we saw, some lineages take the form of a single, unbranching line. But others – interwoven lineages – are built up through splitting and reuniting in an ongoing pattern of interweaving. In developing an historical network analysis, we combine attention to structure characteristic of social network analysis with attention to temporal processes characteristic of historical sociology. The results make it possible to identify larger collectivities that would be invisible from static snapshots. This ability to recognize pattern in historical network data will have wider applicability for the study of phenomenon such as social movements, new industries, or schools of thought in scientific or literary fields noted for emergent categories in fields of relatively low levels of institutionalization.

Thus, as direct counterpart of the topography of intercohesion along the synchronic dimension, we have identified interwoven lineages across the diachronic dimension. In a manner not dissimilar to the intercohesive position – where we find recombination occurring at points of diversity within familiarity – so along the historical dimension, groups form and reform along lines of patterned coherence, separating to encompass greater diversity and rejoining to benefit from familiarity. Whether at the scale of the intercohesive position or along the historical dimension of interwoven lineages, the recombinant work of innovation requires the dual refrain of familiarity and diversity.
Appendix A. The Clique Percolation Method

The clique percolation method operates on clique adjacency. \( K \)-cliques are adjacent if they share \( k \)-1 vertices. A clique of four is adjacent to another clique of four if they share three members. From adjacencies one can assemble a clique chain, traversing along clique adjacencies. The union of all \( k \)-cliques in such a chain form a \( k \)-clique percolation cluster if no more \( k \)-cliques can be added. This contiguous and highly cohesive region of the network is a cohesive group, within which a \( k \)-clique can percolate, or roll along, by always replacing only one of the \( k \) nodes. Using a \( k \) value of four, as we do in this study, yields cohesive groups where all members have ties to at least three other members in the group.

Figure 5. Illustrating the logic of the CPM algorithm.

![Step 1](image1)
![Step 2](image2)
![Step 3](image3)

Figure 5 illustrated the logic of the CPM method in identifying a small cohesive group. We start by identifying a 4-clique, and in step 1 this 4-clique is rolled along by replacing one node. In step 2 the 4-clique is rolled further along, and in the final step, step 3 the group is identified, as there is no further possibility to roll the 4-clique along.

The \( k \) parameter of the CPM method is adjustable. Choosing a lower \( k \) results in a more uneven distribution of group sizes. A \( k \)-value of one is of little utility, since considering the percolation of complete subgraphs of one node means that the whole network is one group. At this extreme value the size distribution of groups is the most extreme – there is just one group that encompasses the whole network. A \( k \) of two means that we consider the percolation of complete subgraphs of two nodes, that is equivalent to considering the percolation of edges. In this case cohesive groups are the disconnected components of the graph. The size distribution of components is very skewed, as a giant component containing an overwhelming proportion of nodes is a common feature of networks. Increasing the value of \( k \) to three means considering triangles (complete triads) as the percolating subgraph. Groups in this case are made of triangles sharing at least two nodes. The distribution of group sizes is more even, although in denser networks the largest triangle-percolation cluster can still be much larger than the second in size. Moving to a \( k \) of four is even more restrictive, since groups need to be dense enough to
allow the percolation of complete subgraphs of four. In this case the distribution of group sizes is more even, and there might not be a clearly largest group.

We decided to use a $k$-value of four in identifying clique percolation clusters. Most applications of the CPM method found that there is a percolation transition between a $k$-value of four and three. While a $k$ of four produces groups that are roughly of equal size, a $k$ of three produces a highly skewed group size distribution. Our data confirms this finding, with a $k$ of three the largest group on the average is three times larger than the second largest, while in some years it is five times larger. With a $k$ of four the largest group on average is 1.21 times larger than the second largest, and the maximal size distance 1.57. Thus we decide to use $k=4$ to identify cohesive groups with the CPM method.

Table 4. Group size distribution and node coverage with the CPM method at various $k$-values.

<table>
<thead>
<tr>
<th>Year</th>
<th>$k=2$</th>
<th>$k=3$</th>
<th>$k=4$</th>
<th>$k=5$</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Largest group</td>
<td>Node coverage</td>
<td>Largest group</td>
<td>Node coverage</td>
</tr>
<tr>
<td>1989</td>
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<td>48.76</td>
<td>1.25</td>
<td>20.85</td>
</tr>
<tr>
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<td>45.23</td>
<td>1.20</td>
<td>25.38</td>
</tr>
<tr>
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<td>57.02</td>
<td>4.73</td>
<td>32.20</td>
</tr>
<tr>
<td>1992</td>
<td>53.30</td>
<td>61.72</td>
<td>4.85</td>
<td>38.83</td>
</tr>
<tr>
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<td>62.92</td>
<td>1.27</td>
<td>42.98</td>
</tr>
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<td>63.13</td>
<td>1.47</td>
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</tr>
<tr>
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<tr>
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<td>54.06</td>
<td>2.11</td>
<td>37.05</td>
</tr>
</tbody>
</table>

Mean 93.11 3.00 1.21 1.17
Min. 18.40 1.20 1.00 1.00
Max. 140.60 5.00 1.57 1.29

Notes: a. Cells indicate the relative size of the largest group compared to the second largest group. b. Cells contain percentages of all firms in the dataset in that given year.
References


