

# The Evolution of Network Structure: Where Do Structural Holes Come From?

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# The Evolution of Network Structure: Where Do Structural Holes Come From?

## **Abstract**

We develop and test a theory of the origins and consequences of structural holes, using a theoretical framework that argues that network structures emerge from the interplay of two complementary forces: structural constraints and network opportunities. Our data comprises a co-membership network among 501 production teams in the Italian TV industry tracked over a period of 12 years. Using analysis that explicitly accounts for endogeneity, we find that structural holes spanned by teams originate from the past status and past centrality, in addition to past structural holes, but are reduced by past team cohesion and past alter content homogeneity, supporting both opportunity exploitation and structural constraint explanations. We also find independent and positive effects of both structural holes and alter content homogeneity on performance.

A great deal of organizational research has focused on the network structural antecedents of favorable outcomes for teams (e.g. Hansen, 1999; Reagans, Zuckerman, and McEvily, 2004), and firms (e.g. Ahuja, 2000a; Baum, Calabrese, and Silverman, 2000). While research on performance outcomes of social structures is valuable, it raises the question of precisely how social structures come about and the processes that shape their evolution over time. Without explicating the logic behind the creation of networks, knowledge of their outcomes remains incomplete because an awareness of the entire chain clarifies the temporal sequencing and causal linkages behind both network emergence and outcomes. A related and more fundamental reason to understand the origin of network structures is the issue of whether they are epiphenomenal or whether they emerge from a set of factors that we can systematically identify and relate to a theoretical model.

Prior explanations for the origins of network structures have typically extrapolated from research on the origins of tie formation which argues that past ties predict future ties, suggesting that structural persistence or inertia shapes the evolution of organizational networks (Walker, Kogut, and Shan 1997; Gulati and Gargiulo, 1999). At the same time, research has also recognized that network structures provide a source of opportunities that help network actors arrive at favorable outcomes. These opportunities, as well as inertial constraints, are not only related to the network's structural characteristics but also its content and nodal properties, all of which the focal actor is exposed to through its network ties (Ahuja, 2000a; Rodan and Galunic, 2004). In this paper we build a theory of the origin of networks arguing that opportunities created by past content and structure give rise to future structures by virtue of opportunity exploitation by network actors (Ahuja, 2000b). More precisely, past networks offer actors a combination of experiences, knowledge access, prominence, and power that can open opportunities and create inducements which in turn influence the evolutionary pattern of network structures. While the past structure gives rise to opportunities, the structural actor avails of those opportunities in a manner that may create a favorable and valuable social structure going forward – subject of course to inertial constraints, also imposed by the very same past structure (Stevenson and Greenberg, 2000).

In sum, we theorize that for focal organizational actors favorable network structures emerge from the combination of the two complementary forces of structural constraints imposed by, and the network opportunities provided, by past network structures and positions. While some scholarly understanding of the factors that influence the formation of relationships between organizational entities exists, in this paper we go beyond the creation of ties to focus on the evolution of structures, specifically structural holes. We combine explanations derived from structural persistence and network opportunities to develop a more holistic and comprehensive theory of the mechanisms and factors underlying the evolutionary dynamics of network structures and their performance consequences.

We focus on the genesis of structural holes because our goal is to better understand the origins of a specific type of network structure rather than that of tie formation or structures in general. Structural holes are present in an actor's network of relationships when the focal actor (or 'ego') is tied to others ('alters') who are not themselves connected (Burt, 1992). Structural holes capture, like other related concepts such as weak ties (Granovetter, 1973; Hansen, 1999), range (Reagans and McEvily, 2003) and brokerage (Xiao and Tsui, 2007; Fleming and Waguespack, 2007), a key network structural property, the efficient and non-redundant access to resources and information.

We investigate these issues in the longitudinal context of 501 TV productions produced and broadcast over a 12-year period in Italy. TV productions are created by temporary project teams, and the teams are interconnected by virtue of industry specialist co-memberships. Our research context of temporary networks, where teams are continuously dissolved and re-created over time, enables us to test the structural and nodal conditions under which favorable network structures arise. A distinctive characteristic of our study is capturing the content of TV productions. Because we aim to explore the origins of structural holes, capturing content allows us to examine how network structure and alter content independently shape the formation of networks in future periods, which also tests an assumption common in the network literature that structural holes reflect content

diversity (Burt, 2004). Moreover, because capable network actors may generate both favorable network structures and positive performance outcomes, in our paper we also tackle the basic issue of the endogeneity of structures and outcomes (Mouw, 2006) by using appropriate estimation techniques (Shaver, 2005).

Our paper makes at least four major contributions. First, we develop and test theory that combines the exploitation of network opportunities provided by past social structures and structural persistence perspectives as factors underlying the creation of structural holes at the organizational level of analysis. In this way, we comprehensively address the factors and processes behind the emergence of network structures. The identification of antecedents of network structures over time also answers the call to expand research on network evolutionary trajectories. Such a research focus is important because “cross-sectional analyses of networks often leave causal relations ambiguous” (Brass et al., 2004: 809). Second, an integrative perspective along the chain that encompasses the drivers of network formation, the resultant network structure, and its effect on outcomes counters the fragmentation of understanding and knowledge about each stage of the process viewed separately (Azevedo, 2002). An articulation of how all the pieces in the causal chain come together also provides the necessary normative link to guide managerial action and behavior. Furthermore, including the performance implications of structural holes confirms the appropriateness of our focus on this specific form of network structure. Third, we contribute to a fuller understanding of the origins of structural holes by including the role of content; specifically, we evaluate whether accessing homogenous production content from its alters might motivate the structural entrepreneur to seek diverse content through structural holes in future networks. Fourth, from a methodological standpoint, high-performing network actors may be better positioned to create superior network structures for themselves which in turn perpetuate superior performance. Such simultaneous co-determination implies endogeneity between structures and outcomes, which we address and resolve by using specialized statistical procedures (Two-Stage Least Squares).

## **NETWORK STRUCTURES AND CONTENT IN CULTURAL INDUSTRIES**

The TV production industry is recognized to be a ‘cultural industry’ in the sense that it produces an aesthetic, symbolic, or expressive product (Lampel, Lant and Shamsie, 2000). Such a product, for example a TV movie, is a creative, non-additive, synthesis of information, ideas and experiences of the specialists that comprise the production team. In the present research, we label the essential symbolic and aesthetic characteristics of a TV production team’s creative output its ‘content.’ As we explain later, in order to identify and capture the key elements of this creative synthesis we develop, in conjunction with industry experts, a detailed set of categories that capture each team’s content.

Moreover, the output of the production team is a prototype in the sense that each product is produced *de novo*. In these situations, product characteristics, including its content, embody and reflect the experiences of the members of the team and are collectively ‘owned’ by them. When these team members move to other teams in later time periods, they carry with them the knowledge, ideas, and experiences accumulated over time from prior teams, like the process of transmission of tacit knowledge across organizational or unit boundaries. Concurrent team memberships work similarly to transmit knowledge across teams.

The production teams in the industry form a large network through interconnected specialists over time. Along these lines, Lampel and his colleagues, referring to temporary projects in cultural industries, note, “The virtue of such latent structures is that they can provide the means whereby a network of specialists that have previously worked together can...efficiently reconstitute the network” (2000: 265). The teams, and the industry network as a whole, can therefore be viewed as a connected universe of identities, values, symbols and artistic expression (Starkey, Barnatt and Tempest, 2000). In consequence, content flows through network ties via the individuals that connect different teams by virtue of co-memberships. In the creation of a TV production, resources accessed in this manner are essential for the enhancement of the creative and idiosyncratic characteristics of the product. Naturally, concurrent links and structures provide different

information to the focal teams than do past links and structures – the latter enable the flow of ideas stemming from experience whereas concurrent links are likely biased toward providing current information. In such co-membership networks, a focal team's alters become defined as those teams on which current team members either serve concurrently (current alters) or have served in the past (past alters). Figure 1 is a representation of the industry as a network of teams which share specialists both concurrently and in the past.

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Figure 1 about here

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Team members adopt a variety of different roles in the team and bring to bear a heterogeneous set of task and skill specialization capabilities on the creation and production of the movie (Baker and Faulkner, 1991). Each team is composed of a director, an assistant director, screenplay writers, the original author, actors, music creators, the producer, one or more executive producers and so on. Thus, every team is by and large required to possess the entire range of skills needed to produce a TV movie, depending on the kind of movie being produced. Except for small variations, heterogeneity in skills and roles is limited across teams, but more importantly does not substitute for the heterogeneity of new ideas, working processes, routines and so on which are accessed from experience on other concurrent and past teams.

It is important to understand that for the production teams, the causality between actions and outcomes on both economic and artistic dimensions is ambiguous, resulting in a high degree of uncertainty about the likely success of the ultimate product (Holbrook and Hirschman, 1982). The uncertainty is only resolved after the production is created, produced and broadcast. Until that time the production team faces a number of decisions regarding the process of combining multiple identities and experiences; capturing the *zeitgeist*; diverging from or conforming to the dominant genre, theme, or content; and dealing with a key challenge of successful cultural products, that of

combining the imperatives of efficiency and creativity (Hirsch, 2000). Performance in this industry, and the ultimate dependent variable in our study, is the eventual commercial success of the production. We move now to a discussion of the theoretical issues around the origins and role of structural holes spanned by the focal team in our network of interconnected teams.

## **THEORY**

The structural holes perspective has attracted considerable interest because they are considered a form of valuable ‘social capital’ (Adler and Kwon, 2002), and thereby present a social structural antecedent for many kinds of individual, team and organizational outcomes. Further back in the causal chain, two fundamental explanations can be identified underlying the creation of social structures: the opportunities inherent in prior networks which may enable an actor to create or re-create future structures, and the inertial constraints imposed by prior network structures themselves (Sewell, 1992; White, 1992). Constraints and network opportunities thus parallel Giddens’ (1984) conception of the duality of structure and action as acting and interacting in ways that mutually reinforce and perpetuate social structure through a ‘structuration’ process (Sydow and Windeler, 1998).

More precisely, the opportunities provided by networks are exploited in future periods by virtue of two mechanisms: one deriving from purposive action and the other due to a less teleological process. First, opportunity exploitation can be considered as purposive action which may create structures by forming or dissolving network links, somewhat evocative of Child’s (1972) notion of strategic choice. Network actors exploit opportunities arising from past patterns of behavior which lead to experiences and knowledge that in turn motivate and may enable an actor to exploit, recreate, and re-configure past network positions into future beneficial ones. In this vein, Burt’s conception of structural holes as social capital highlights the entrepreneurial role of the network actor in the generation of this valuable form of social structure (Burt, 1992).



Second, a less directly teleological mechanism may also play a role in generating networks. Specifically, positions in the past networks can provide focal actors with opportunities that can shape future networks independent of actors' ability or intention to strategically exploit past positions. For instance, adopting a similar logic, Powell and his colleagues suggest that central and high status actors are likely to receive a disproportionate share of future ties, referring to this network evolutionary process as 'accumulative advantage' (2005: 1140). This less teleological mechanism can amplify future changes in the structural characteristics of past networks by reinforcing the brokerage position of prominent actors over time (Fleming and Waguespack, 2007).

Conversely, rather than the opportunities that actors can exploit and shape, their current set of interactions produce social structures that tend to persist and reproduce themselves over time through norms, rules and social pressures. In turn, this process creates inertial forces that shape and constrain an actor's behavior over time (Parsons, 1951). This structural explanation suggests a strong element of stability and path dependence, and emphasizes the role of inertia and relational lock-in in network dynamics. Relatedly, some scholars consider the formation of social structures as the result of "...individual members' disciplined compliance with group expectations" (Portes and Sensenbrenner, 1993: 1325), implying again that it is strongly affected by previous structures and ties (Madhavan, Koka and Prescott, 1998).

In this paper, our goal is to build a theory around the interplay of the exploitation of network opportunities and the role of structural persistence as forces behind the emergence of structural holes. Within this framework, we employ both structural and non-structural (or nodal) predictors. The former include, for a focal team, past structural holes, past team cohesion (prior working relationships among team members) and past centrality which together represent the dominant set of structural properties driving outcomes in social network research (Coleman, 1988; Burt, 1992; Adler and Kwon, 2002; Tsai, 2002). The non-structural set includes team status derived from links with successful past alters, and past alter content homogeneity (similarity of past alter content), which comprise another class of explanations for network-driven behaviors and outcomes,

particularly in uncertain conditions (Podolny, 1993). We now integrate the factors we investigate as antecedents of the formation of structural holes within our theoretical framework.

In our research context, due to the temporary nature of the network, the reconstruction by the focal team of the structural holes that team members were spanning in the past is a manifestation of the exploitation of past network structures. Put another way, in this sphere structural entrepreneurs exploit past network opportunities by reactivating structural patterns that were valuable in the past. Next, the content diversity to which actors were exposed by their past links offers focal actors the inducement to seek a more diversified set of alters through structural holes by reconfiguring past schemas. Further, past centrality and status derived from past alter performance give actors the opportunity to exercise their judgment to choose amongst potential future team members in a way that reinforces or enhances their favorable structural position. Status orderings and structural popularity (centrality), as Podolny (1994) points out, become particularly valuable signals under uncertain conditions, such as those that exist in our research context. This element is reflected in the roles played by our prominence variables of team status derived from past alter performance and past team centrality.

We now move on to the second major dimension of our framework: inertial constraints or persistence. Persistence is the extent to which interactions are reproduced over time and across a number of actors who develop what Giddens refers to as ‘structural properties’ or institutionalized frameworks that are reproduced across time and space (Giddens, 1984). Thus, persistence is not driven by teleological behaviors, judgment, or autonomous reconfiguration but rather by a process that is subject to the inertial, constraining effects of prior patterns of relationships. In our framework, the variable that typifies the notion of persistence is team cohesion. Team cohesion is an expression of accumulated past working relationships that constrains the ability, motivation and preferences of individual actors toward preserving past patterns. Teams with high cohesion (many prior working relationships) in the past will tend to find themselves in tightly-linked structures in subsequent periods because future teams that employ the cohesive members of a prior team will

tend to replicate previous connections. Such actions will result in fewer structural holes for the focal team.

In sum, our theory is built on two planks: in network opportunity exploitation and structural persistence it identifies the basic explanations for the creation of social structures. Below, we proceed to develop specific hypotheses.

### **Structural Holes and Past Alter Content Homogeneity**

It is a truism that networks give to actors access to alters' experiences, ideas, information and knowledge (Gulati, Nohria and Zaheer, 2000; Ahuja 2000b). We argue that homogeneity and similarity of content among past alters provide the motivation for the core membership of the team, which is composed of its key roles, to seek out diversity in the content of future alters. This argument, drawing on the network opportunity exploitation logic, adopts a strategic choice approach which implies that the team's core membership may reconfigure its network to span structural holes in future networks with the expectation of accessing the requisite diversity to avoid the problems stemming from high levels of content homogeneity among past alters.

Our argument also builds on the well-accepted argument in the research on structural holes that views content and structure as mirror images of each other. Thus, structural holes may be created in order to reach out to alters that hold diverse and novel content. Burt makes this argument explicit in a recent paper. As he writes, "The presumption...is that the content of ideas reflects the social structure in which they emerge" (2004: 5). At the same time, although disagreeing over which structure is more beneficial, Coleman (1988) also links structure and content, and argues that network closure, implied by few structural holes, is associated with increasing similarity and conformity among actors. Our search for the antecedents of structure also enables us to test the logic underlying this link. We do so by not only splitting apart content from network structure but also investigating the causal relationship between prior content and subsequent structure.

In our research context, because novelty and innovation are the keys to success in the industry, homogeneity among the project teams from which its specialists are drawn is likely to hamper the focal actor's ability to develop a successful production. Homogeneity may also be engendered by the clumping together of genre-driven communities. Common mental models, groupthink, and unproductive lock-in to sterile ideas will begin to hurt a focal team's creative potential (Janis, 1972). Thus, actors' exposure to similar content create an unfavorable context which induces the activation of a network reconfiguration mechanism seeking for a more diversified set of alters. Consequently we hypothesize:

*H1a: The higher the content homogeneity among past alters, the higher the structural holes spanned by the focal team in the present.*

Conversely, an alternative perspective may reach the opposite conclusion about the constraints imposed by alter content homogeneity. Under such conditions of alter content homogeneity, routines and standard operating procedures may develop more easily within the focal team thereby enhancing efficiency (Cyert and March, 1963). Homogeneity may also enhance outcomes via a more accurate understanding of the skills and capabilities that other members of the project team might possess, and what has been referred to as knowing who-knows-what, or 'transactive memory' in the literature, may improve coordination and limit inefficient duplication for the production task at hand (Wegner, 1986). When the focal team is locked-in with a group of alters with similar content, it may enhance its proclivity to seek out similar alters in the future. Because of homophily, these alters may be connected to each other, thereby reducing structural holes in the following period for the focal team. Therefore,

*H1b: The higher the content homogeneity among past alters, the lower the structural holes spanned by the focal team in the present.*

## **Structural Holes, Status and Past Centrality**

Teams may seek to form connections with high-performing teams to signal their status to the market (Podolny, 1993; Zuckerman, 1999, 2000). More specifically, the status of a focal actor derives from the performance of the *alters* with whom it is affiliated (Benjamin and Podolny, 1999). Scholars have distinguished status from economic notion of reputation, the latter being tied much more directly to the performance of the focal actor itself (Shapiro, 1983; Washington and Zajac, 2005). Such a signal of high status is particularly valuable when market uncertainty is high, which characterizes our context well because other cues about the inherent quality of the team are missing (Podolny, 1993).

The relationship between status and structural holes has been addressed by Podolny who argues that it is reasonable to hypothesize a high correlation between the two constructs because an actor with structural holes is by definition prominent in the overall network. However, he goes on to suggest the existence of "...a real trade-off between the formation of ties that will add structural holes to the network and ties that augment the actor's status" (2005: 233) because reaching out to lower status players may increase structural holes but reduce status, implying a negative relationship between status and structural holes. From a different perspective, Burt (1992) has argued that a network position rich in structural holes, by which an actor is connected to a large number of disconnected alters, might be beneficial in establishing a positive reputation. Thus, given the ambiguity in the literature, both the direction of the relationship between status and structural holes and the possible causality between them is important to assess.

Broadly, the status that current teams inherit from the performance of past alters and the prominence gained in past social structure give teams the opportunity to exercise selection power over current alters. We argue for a positive relationship between team status derived from past alter performance and structural holes that arises from at least three mechanisms. First, high status teams prefer disconnected alters because teams are reluctant to embed themselves in tightly-connected networks where the risk of knowledge spillovers among cliques would be higher than in open

networks. Since tightly-coupled networks create generalized access among their members to such spillovers, high status teams will prefer disconnected structures in order to maintain more effective control over these knowledge and information resources through brokering. As well, since the products in our industry are creative and artistic combinations of ideas, the high status focal team may also worry that highly-connected alters could ‘gang up’ against it, and therefore again prefer a sparse network structure with plentiful structural holes. The well-established control benefits of a network broker play directly into this argument.

A second, and related, explanation for the creation of sparse networks by high status teams is that when approaching potentially new ties, the focal high status team may be able to require exclusivity, thereby reducing the risk of knowledge spillovers from the high status focal team. Such exclusivity would also imply more structural disconnectedness in the network. Of course it is possible that low-status actors may have to pay a premium to join a high-status team, which may limit the availability of low-status actors, but the social rank conferred by high-status teams, especially in a cultural industry context, may make this cost worthwhile.

A third, less directly teleological mechanism may also play a role in generating networks with plentiful structural holes for high status focal teams. This mechanism essentially operates by virtue of the popularity of high status teams which employ well-known and successful specialists. Members of low status teams will prefer to work in high status teams simply because of its fame, insights and knowledge, rather than because of common alters. Therefore the likelihood that the low status teams will be connected with each other is lower, resulting in structural holes among them. From this viewpoint, current structural holes in focal team networks are the outcome of high status focal teams accepting disconnected specialists into their teams. This is a mechanism that relies, in addition to focal team opportunity exploitation, on autonomous or ‘collateral’ effects of alter behavior; although a factor operating in the opposite direction is that a limited number of high-status teams may force low status teams to connect with one another. At the same time, high status

teams that comprise successful people from past teams are also in demand from different venues in the industry, which also would beget structural holes for the high-status focal team.

A similar set of arguments apply when the focal actor has been connected to central players in the past. Centrality, which has been shown to be related to performance (Tsai and Ghoshal, 1998), imparts prominence to a team and can both enhance the power of the focal team to demand exclusivity, as well as attract disconnected others to the focal team thus creating structural holes in a later period. An alternative explanation is that teams that have had a higher propensity to form ties in the past may be composed of individuals who are part of a team that is also similarly highly inclined to form ties in the current period. Thus, in addition to viewing status of focal project teams as based on alters' past success, we also conceptualize the prominence of the team in purely network terms as its centrality. Accordingly, we hypothesize that:

*H2a: The higher the status of the focal team, the higher the structural holes spanned by the focal team in the present.*

*H2b: The higher the past centrality of the focal team, the higher the structural holes spanned by the focal team in the present.*

### **Structural Holes, Past and Current**

One of the more fundamental themes in the literature on structural holes is the notion that the actor spanning structural holes gains brokerage and control benefits from its position. Burt describes control as “giving certain players an advantage in negotiating their relationships” (1992: 76). Expanding on this theme, Burt invokes the metaphor of *tertius gaudens*, originally due to Simmel (1922), as the ‘happy third’ that is able to play off one alter against another – although perhaps a more apt expression in this situation might be *divide et impera* (Krackhardt, 1998). The latter focuses on the arbitrage benefits of brokerage and control, which are well-recognized as deriving from the exploitation of the information asymmetry between alters that span the structural hole. A number of other literatures echo the notion of advantage from asymmetric positions,

including resource dependence theory (Pfeffer and Salancik, 1978) and transaction cost theory (Williamson, 1985).

Our argument here is that the actor bridging structural holes in the past may exploit opportunities to recreate them in order to maintain the asymmetry embodied in the position to gain brokerage and control benefits (White, 1992). Thus, over time, actors may endeavor to replicate their privileged position, using the social capital accumulated from their past relations (Pollock, Porac and Wade, 2004). However, while the specific holes in past structures will vanish over time with the dissolution of both the teams and the network, we argue that the core membership of the focal team may use power to recreate holes in the team's current social structure.

In our context, the industry is in a state of continuous evolution due to the inherent nature of temporary projects which culminate in TV productions. In addition, industry boundaries are unstable. These factors create conditions which give structural entrepreneurs and brokers numerous opportunities to construct and reconstruct their social structures. Being able to connect with other production teams which are disconnected from each other gives the focal team a broker's control over flows of information and knowledge across teams. We formalize the foregoing arguments by predicting that past structural holes in the network of the focal team will be reproduced in current social structures. Therefore,

*H3: The higher the structural holes in the past, the higher the structural holes spanned by the focal team in the present.*

### **Structural Holes and Past Team Cohesion**

Another element of the genesis of structural holes lies in the past structural embeddedness of relations. Our theorizing in this regard draws on the notion of inertial constraints or structural persistence. We develop arguments predicting that past team cohesion will persist over time reducing the likelihood that structural holes will be generated in a future period. Past team cohesion refers to the extent to which members of the current organization or team have worked together in the past. When such social bonds



are created, they resist rupture, and the persistence that they manifest translates into connections between teams in later periods. In turn, such bonds generate more cross-cutting ties among alters and therefore fewer structural holes in the focal team's network.

The persistence of social structures has been a common theme in the sociological literature (e.g. Suitor, Wellman, and Morgan, 1997). Scholars have argued that social structures are created as the results of "...individual members' disciplined compliance with group expectations" (Portes and Sensenbrenner, 1993: 1325). This view suggests that social structures are formed through the reproduction of norms and behaviors embedded in past social structures (Giddens, 1984). Social structures are therefore subject to inertia and path dependence, and represent stable, institutionalized patterns of relationships (Bourdieu, 1986).

Considerable research finds empirical support for the idea that social structures reproduce themselves. Moreover, individuals and organizations are more likely to enter into new relationships the more relationships they start with. Gulati (1995) demonstrates that the "social context resulting from cumulative prior alliances influences [subsequent] alliance formation" (1995: 643). Similarly, Walker, Kogut, and Shan (1997) find that patterns of interfirm relations tend to persist over time. Ties may be repeated and consequently become stronger and more durable; time nurtures and cements relations. Overall, these arguments suggest that social structures tend to persist over time. Thus, a structure with high levels of internal cohesion generates norms, trust, obligations and reciprocity, all of which impede and constrain its ability to change over time.

We argue that network structures with a high degree of within-team cohesion resist pressures to get converted into structures with numerous structural holes in later periods. Thus, cohesion implies that teams are more easily able to reach out and connect to teams with specialists who have worked with their current members in the past than with those that have not. Specifically, future teams that employ the cohesive members of a prior team will tend to make connections between scattered prior team members, resulting in fewer structural holes and conversely, greater

closure amongst them. This effect may also work through team members persuading teams to hire prior co-workers to maintain and enhance the social bonds amongst one another.

In sum, the effect of within-team cohesion on the genesis of structural holes among teams is created by virtue of structural persistence. Moreover, due to the nature of cultural products, identity is a crucial element that coalesces and amplifies the persistence of cohesive structures over time. As well, teams themselves may recognize the efficiency value of shared language and routines and may prefer to hire workers that have worked together in the past. These individuals manifest co-memberships through fewer structural holes in the focal team's network because they are also concurrent members of other connected teams. Formally,

*H4: The higher the past team cohesion, the lower the structural holes spanned by the focal team in the present.*

## **THE CONSEQUENCES OF STRUCTURAL HOLES**

Since the aim of the paper is to investigate in an integrative manner the entire causal chain from the antecedents to the performance consequences of structural holes, the second part of our theoretical model relates structural holes to performance, together with current period alter content homogeneity. We combine these two predictors because we wish to understand whether structural holes still influence performance after alter content is disentangled from structure and the endogeneity of structure with performance is taken into account.

### **Alter Content Homogeneity and Performance**

We extend our reasoning to make the connection between concurrent alter homogeneity and performance in the current time period. Two major and opposing lines of thinking may be identified regarding the relationship between diversity or homogeneity and performance. The first is rooted in the notion that diversity and variation are beneficial for the performance of teams because they strengthen the ability of team to deal with uncertainty, complexity, and non-additive

problems. Diversity helps the team in reducing the risk of groupthink and avoiding cognitive traps (Janis, 1972; Kahnemann and Tversky, 1979). Our argument here builds on the structural holes perspective which argues structural holes enhance performance because they capture diversity and novelty in ideas by tapping into alters that are disconnected with each other (Burt, 1992).

Conversely, low diversity, or high alter content homogeneity, may actually produce heightened quality, as we argued earlier. In situations where past alters are homogenous in their content, it becomes easier for shared specialists to transfer skills, routines, and knowledge to the focal team because of content similarity in the alters' creative products. Superior outcomes may result from access to alters with similar content because of the efficiency inherent in absorbing and applying skills and knowledge which share a common base (Cohen and Levinthal, 1990). Moreover, homogeneity may benefit the process of production because of a better understanding of the capabilities of team members, also referred to as who-knows-what, or 'transactive memory' (Wegner, 1986). Research shows that show that team diversity may hurt performance and even though creativity may be enhanced, eventual implementation success is negatively affected (Ancona and Caldwell, 1992).

In the context of TV production, a production team with high alter content homogeneity implies that the focal team is exposed through the network to specialists who are currently working in teams with similar content. As several scholars have pointed out, success in this industry requires a combination of creativity and efficiency (Lampel, Lant and Shamsie, 2000). The potential downside of such similarity is that the creative process may be stifled by the lack of new and different ideas and opinions, and because creativity is an important element of a successful product, performance may be hindered. On the other hand, the quality and knowledge-related creativity of the project team's processes may be enhanced, resulting in higher performance. Reflecting the strong opposing arguments, we propose two competing hypotheses for the effect of alter content homogeneity on team performance:

*H5a: The higher the homogeneity of current alter content, the higher the performance of the focal team.*

*H5b: The higher the homogeneity of current alter content, the lower the performance of the focal team.*

## **Structural Holes and Performance**

Our reasoning here links structural holes to performance. Even after we tease out the role of diversity, the benefits of structural holes arguably operate through mechanisms of control, brokerage and the exploitation of information asymmetries between disconnected alters. In our setting, the project teams are in competition with each other for favorable time slots, channels, and viewership. We suggest that structural holes might still provide the focal team here with the power of arbitrage and competitive intelligence in regard to information that is not merely based on content. Examples include production costs and plans for placement with preferred channels. Knowledge of these, particularly from several alters, might enhance the ability of the focal team to arbitrage its knowledge to improve its position vis-à-vis its disconnected alters. Moreover, even if disconnected alters are similar in terms of content, knowledge, ideas and experience, a specialization effect may yet aid the focal team. This may take the form of allowing the focal team to fine-tune its product and processes by competitively exploiting production or market niches that are unoccupied by other teams.

To summarize, although we have argued previously that there may be a positive effect on performance from efficient routines that alters with homogenous content may help create, we argue that bridging structural holes in the external network helps the production succeed in one of two ways. First, from the perspective of a focal team, connections with other disconnected teams may provide access to novelty and diversity in skills, experiences, ideas, capabilities and resources resident in the other teams. Second, if the value of efficient routines from homogenous alter content

trumps the value of novelty and idea diversity, the control benefits of spanning structural holes may yet yield a positive performance effect for the focal team. Thus, we hypothesize:

*H6: The higher the current structural holes in the network of the focal team, the higher its performance.*

## **METHODS**

### **Reconciling Network Theory across Levels of Analysis**

The sizeable research on co-membership networks among boards, clubs, movie productions and so on has implicitly treated the network of ties between nodes as isomorphic with individual networks (Mizruchi, 1996; Zajac and Westphal, 1996; Haunschild and Beckman, 1998). Before we present our data we need to clarify the conditions under which research on group or team co-memberships can in fact be regarded as isomorphic with research on individuals as network nodes.

As we see it, from the perspective of theory, at least three implicit assumptions underlie a shift in the level of analysis from networks of individual-level nodes to higher levels of analysis (say, teams). The first is what might be called the assumption of *composition*; the second is what we refer to as the assumption of *contagion*; and the third, the assumption of *causality*. We explicate each in turn and thereafter explain specifically how our research considers and addresses each of the three key assumptions – of composition, contagion, and causality – for validly porting theory from the individual dyadic perspective to that of the collective actor.

The *composition* assumption tends to assume that a tie between two teams through a single link connecting a part of one team to a part of another team represents a link between the two teams as a whole. While clearly valid at the individual level of analysis, when the node is a single person, the underlying logic needs additional justification at higher levels of analysis. For example, the research on board interlock networks typically assumes that the co-membership link between the boards of two organizations connects the organizations themselves and influences the actions of the organizations as a whole. However, close intra-organizational interactions, linkages and

communication processes need to be in place for the assumption of composition to hold at the higher level of analysis.

The teams in our industry, as we illustrate in Figure 1, do form tightly-coupled networks *within* teams. In the same vein, Uzzi and Spiro's (2005) Broadway musical teams are considered "fully-linked cliques." Consequently, when two teams share a specialist, because coordination processes are so tightly coupled, all the members of the team are influenced by the link and the co-membership relationship between the two teams in fact becomes a knowledge and experience conduit for the team as a whole.

The second assumption is that of *contagion*. Network research at the individual level is able to assume with some justification that network content flows through individual nodes to other nodes which are not linked directly to each other (i.e. content passes from X to Z through Y even though X and Z are not directly linked). Classic research on the diffusion of ideas through networks (e.g. Coleman, Katz and Mendel, 1966) illustrates this phenomenon at the individual level of analysis. Again, at higher levels of analysis, it is problematic to automatically believe that a contagion process exists. In order to make an assumption of contagion, researchers should ideally specify theoretically or empirically, or both, the pathways through which content moves through to the indirectly connected organizational node.

In our case, as in the earlier example when two teams X and Y share a specialist (say A) and another specialist (say B) is shared between Y and Z, contagion implies that content passes between X and Z through Y. In this case again, due to the tightly-coupled nature of the team, content is likely to flow through a contagion process. However, we note that, compared to a network of individuals, even in tightly-coupled teams the contagion process may be diluted because the process is necessarily mediated through coordination and communication interfaces within the team. On the other hand, the moderation of the contagion processes may amplify the brokerage power of Team Y. Structural holes in such co-membership networks may therefore be an even more potent source of explanation.

The third assumption, that of *causality*, focuses on the distance between cause and effect. In a network of individuals, it is easier to identify the causal chain between structural cause and effect because processes leading to individual-level behavior and outcomes can be narrowly circumscribed. However, when the node is a collective and complex actor, the causal chain is harder to tease out and the process through which alliance structural content translates into firm-level performance is tenuous and rarely, if ever, specified. It is important, then, to draw out the causal chains when the network involves higher levels of analysis. However, the question of causality should also consider two boundary conditions. First, causal reasoning has to jointly take into account the type of tie with the type of the outcome. For example, the causal relationship between R&D linkages among biotechnology firms and firm innovation is clearly more direct than that between email networks and promotion at the individual level. Second, the relationship between the boundary spanner and the organization as a whole may also influence the strength of the causal mechanisms.

The teams in our research are, as mentioned before, tightly coupled. At the same time, critical roles also exist in the team, such as those of the producer and the director, to which intentionality can be ascribed. These characteristics, tight coupling and existence of critical roles, help justify the actions of a team in our research as one that behaves as a unitary actor helping satisfy the causality assumption. Finally, while these three assumptions when satisfied allow researchers to apply network theory across level of analysis, they are not unrelated. More precisely, contagion and composition maybe viewed as substitutable in the sense that if either condition is fully satisfied it makes the second unnecessary.

### **Data and Model Lag Structure**

We test our hypotheses about the genesis and outcomes of structural holes by studying TV productions in Italy over the period 1988-1999. Our dataset includes all TV productions (TV movies, serials, and so on) produced and broadcast by any of the six national TV channels (which

cover about the 95% of the global TV audience in Italy) during this period. We collected three kinds of data. First, we gathered longitudinal data on teams, all their members, and their networks of relations from the annual reports of TV movies and serials in Italy published by the state-owned broadcaster RAI, which includes productions broadcast by all channels. Second, from the appendix of this publication we gathered detailed synopses of each TV production. Third, we collected audience data for TV production teams from Auditel, an independent agency appointed to measure the actual viewership of each production. We excluded rebroadcast audience numbers in order to give all the productions an equal chance of reaching an audience, because older productions have a much higher probability of being rebroadcast, and we wanted to use a consistent measure of performance success as the audience number from the initial broadcast. In sum, the dataset contains information on all the 4793 specialists that participated in all the 501 television productions created and broadcast over that period.

Because the theoretical purpose of the paper is to surface the mechanisms and antecedents which underlie the genesis of network structures, we needed a long enough history or ‘past’ window. Thus, we split the sample at the median of the data (1995) which gave us five years of current observations (1995-1999) and seven years of past data for each focal team (1988-1994). With such a split, we obtained 249 focal teams, for which we computed our endogenous variable of current structural holes. The remaining 252 productions from 1988-1994 were used to compute measures of the past. To measure these latter variables we used a time window of seven years, which corresponds to the ‘longest past’ we could obtain using 1995 as a cutoff. By moving the seven-year window across multiple years (five times, corresponding to each of the five years 1995-99) we captured the same time span (or past) for all productions broadcast in the period 1995-1999 (e.g. for a production broadcast in 1996 we used past network data for 1989-1995 and so on).

Thus, past and the current variables are based on different periods of the data – the past on the seven years preceding the focal team production year and the current on the year of production and past and present measures do not share any overlapping years of network data. Moreover, it is



important to remember that the teams themselves did not exist in the past and to that extent there is no scope for any kind of ‘fixed effect’ or tendency of the team to exhibit autocorrelated errors over time. The lag structure adopted to develop measures of antecedent variables for the seven-year past takes the following form:

$$y_t = \beta_0 + \beta_1 \sum_{i=1}^7 x_{t-i} + \dots + \varepsilon_t$$

### **Analysis and econometric approach**

We used a 2SLS model with a robust variance estimator to control for the effects of correlation between errors across equations due to endogeneity between network structure and performance. Although Baron and Kenny (1986) recommend the use of 2SLS only for controlling possible reverse causality from the outcome to the mediator, Shaver has recently suggested that 2SLS “...is an effective estimation strategy in a much broader set of circumstances...even when feedback is not a concern” (2005:339). He recommends its use because of the power of the methodology to handle potential correlation among error terms in the equations. The 2SLS procedure takes into account such correlations and produces coefficients that are consistent and unbiased.

Further, because our dependent variable of performance (Audience Share) is bounded, we adopted a *Tobit Two-Stage Least Squares* analysis (tobit with endogenous covariates in Stata, or *ivtobit*) that provides more consistent estimates in this case than two-stage least squares without a tobit specification. Although our endogenous measure of current structural holes (Efficiency) is bounded [0-1] as well, Angrist and Krueger (2001) point out that in a two-stage procedure, it is not necessary to use limited dependent variable estimation for the first stage, even if the endogenous variable is bounded, to generate consistent estimates in the second stage. Nevertheless, because tobit with endogenous covariates analysis estimates the second stage as a tobit model and uses OLS in the first stage, in order to check the consistency of our first-stage estimators, we also ran a series

of tobit models for the first stage alone (our instrumental variables) to account for the bounded nature of our endogenous variable (structural holes as efficiency).

We checked the consistency of and the appropriateness of the 2SLS modeling approach with several tests. We began with the Wu-Hausman F-test and the Durbin-Wu-Hausman  $\chi^2$  tests. These are tests for endogeneity where the null hypothesis states that an ordinary least squares (OLS) estimator of the equation would yield consistent estimates, and thus endogeneity among the regressors would not have deleterious effects on OLS estimates. Moreover, due to our use of a large number of instrumental variables, we also checked for the presence of overidentification of our model with a Sargan test, which provides a measure of instrument relevance for all instruments. Inability to reject the null hypothesis indicates that the model is not over-identified and is acceptable for the two-stage procedure that we are employing.

### **Analysis of TV production content**

Content analysis is a systematic, replicable technique for condensing a large number of words of text into content categories based on a set of explicit rules of coding (Berelson, 1952; Krippendorff, 1980; Weber, 1990; Stemler, 2001). Holsti offers a broad definition of content analysis as "any technique for making inferences by objectively and systematically identifying specified characteristics of messages" (1969: 14). Content analysis can be a powerful tool for determining artistic identity because authors synthesize it, as well as the messages they wish to communicate, through their writing. In cultural contexts, when the artistic result is largely based on teamwork, the script becomes the crucial document for sharing, communicating and understanding the team's meaning and identity. We adopted an *a priori* coding procedure where the categories were established prior to the analysis based on recommendations by industry experts. The source for team content is a synopsis of the script (three pages on average) developed by the production team for the national archive of TV productions.

We used the following procedure to analyze the productions' content. First, we independently reviewed the scripts and arrived at a set of 19 content categories. Second, we invited a panel of six industry experts to validate our list of categories and variables. Based on input from this panel, we pared down the initial list of 19 categories to 12 (see Appendix and Table 4 for details of categories and variables). Third, two researchers utilized the final checklist to independently code production content (1 if the content of the production is consistent with a variable, 0 otherwise). Fourth, we checked for inter-rater reliability across variables using Cohen's kappa (Cohen, 1960) to measure the degree of agreement between the raters. The K value may be interpreted as the proportion of agreement between raters after accounting for probability (Cohen, 1960). If the two initial raters did not agree, a third rater repeated the previous steps and the value chosen was that of the majority of raters. The kappa coefficient (K) for the overall reliability of our 12 content variables was .80. This value compares favorably with the literature using kappa which suggests that a coefficient of .61 represents reasonably good overall agreement (Kvalseth, 1989).

### **Two-Stage Least Square Analysis (2SLS) with Tobit: First Stage Variables**

**Endogenous variable: Current Structural Holes.** We measure structural holes as the efficiency index in the network of current ties among production teams. We use Burt's (1992) measure of efficiency that counts the ratio of non-redundant ties to total ties for a focal team as

$$\left[ \sum_j \left( 1 - \sum_q p_{iq} m_{jq} \right) \right] / C_j$$

where  $p_{iq}$  is the proportion of the focal TV production team  $i$ 's ties in connection with team  $q$ ,  $m_{jq}$  is the marginal strength<sup>1</sup> of the relationship between team  $j$  and team  $q$ , and  $C_j$  is the total number of

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<sup>1</sup>  $M_{jq}$  is the marginal strength of relations from contact  $j$  to  $q$  (Burt, 1992). It expresses the ratio between the interaction of team  $j$  with team  $q$  divided by the strongest of  $j$ 's relations with any other team. Formally we have:

$$M_{jq} = (Z_{jq} + Z_{qj}) / \text{MAX}(Z_{jk} + Z_{kj})$$

ties for team  $i$ . A high value of efficiency for team  $i$  indicates that its ego network is non-redundant and thus rich in structural holes. This measure captures the non redundancy of  $i$ 's ties as the degree to which a focal team  $i$  has many independent ties. More specifically, this measure estimates the degree to which  $q$  is a large proportion of  $j$ 's ties, and  $i$  has ties with  $j$ .

### **Instrumental Variables**

**Past Network Variables.** To compute past network variables, as we explained earlier, we use seven-year moving windows. As a specific example, let us consider the computation of past structural holes of the team number 273 produced in 1995, which uses as a 'past' all the 252 productions produced in the seven-year period 1988-94. To measure the past structural holes of this 1995 team: 1) We began with an input dataset of all ties among all industry specialists in the past time window 1988-1994, which is a  $4793 \times 252$  matrix of 252 vectors, each representing a team from the past with all 4793 individual specialists where  $x_{ij}$  equals '1' when specialist  $i$  is part of team  $j$  and '0' otherwise; 2) We then created a vector of size  $4793 \times 1$  for the focal team number 273 (produced in 1995); 3) Next, we joined this vector to the first matrix creating a new matrix sized  $4793 \times 253$  which now included all the potential past alters for team 273; 4) We then 'affiliated' this latter matrix to make it a co-membership team-by-team matrix of size  $253 \times 253$  where  $x_{ij}$  is a count of the number of specialists shared between team  $i$  and team  $j$  (note that in the analysis we control for team size); 5) On this co-membership matrix we calculated network measures (e.g. past structural holes) for team 273; 6) Finally, we repeated this procedure for all 249 current focal teams that comprise our dataset (the set of production teams with 'pasts').

**Past Structural Holes.** By applying the procedure described above and adopting the same efficiency measure we use for Current Structural Holes, we measured Past Structural Holes as the ratio of past non-redundant ties to total past ties for each focal team.

**Past Team Centrality.** Past team centrality is measured as the Freeman degree centrality of focal team in the network of past ties (over a seven-year window).

**Status.** Consistently with Podolny's (2005) conception of status as connections with high-performing alters, we measured the status of the focal team as the accumulation of the past performance of past alters, which we standardized to correct for its skewed distribution. In our context, performance in terms of audience numbers and social order are strongly correlated because of the highly socially-constructed nature of cultural industry performance. As well, this success flows through affiliation via shared co-membership links over time, which is consistent with a sociological conception of status. Thus, it is the success of the prior team that is being carried forward through an associational link – which in our case just happens to be through a shared membership in the teams – and which confers status upon the focal team. Moreover, because teams are composed of several specialists, including those who perform technical or lower-level tasks and are not involved in the creative expression of the production, we only selected the alters linked to the focal team by critical roles in TV production. These include: the director, screenplay writers, original author, producer, and actors in major starring roles (average: six per team). To account for the decay effects of status over time we weighted the more recent successes more by using a decay function based on the age of the past alters' broadcast (i.e. 1/7, 1/6 and so on).

Given a focal TV production team  $i$  at time  $t$  and its  $m$  past alters, formally we have:

$$S_i = \sum_{j=1}^m (P_{j_n} / n)$$

Where:

$P$  is the performance of  $j = [1, m]$  past alters of focal team  $i$ ;

$n$  is the time lag =  $[1, 7]$  between focal team  $i$  and past alters' team  $j$ .

**Past Alter Content Homogeneity.** In order to measure the homogeneity among past alters we measured the content similarity among the past alters of each focal production teams from the

content analysis. We used the 12 content variables described in the Appendix and Table 4 to assess the contents of the 501 TV productions in our dataset. We transformed the two-mode matrix of Production by Content (with dimensionality  $501 * 12$ ) into a one-mode Production by Production matrix where  $x_{ij}$  is the degree of content homogeneity among productions  $i$  and  $j$ . To do so, we used the similarity procedure of UCINET VI and adopted the measure of similarity as the proportion of exact matches that computes the proportion of cases in which  $x_i = y_i$  for all  $i$  (Borgatti, Everett, and Freeman, 2002).

**Past Team Cohesion.** Past team cohesion refers to the density of relations *among* the members of a team. The measure we use computes cohesion as the valued density of past ties among the members of focal teams. Specifically, the measure captures for current members of a focal team their previous collaborations over the prior seven years – when they were working together in the past. For a valued network like ours, each prior tie is weighted by the number of previous collaborations.

We illustrate this procedure with an example. Consider the computation of past density (cohesion) for the focal team number 489 produced in 1999. To measure the past density of this 1999 team the datasets we used were: 1) the input dataset, a  $4793*4793$  matrix containing all relations among all specialists in the past time window 1992-1998; and 2) a ‘blocking’ dataset of size  $4793*249$ , essentially an affiliation matrix, where a value of 1 indicates when a given specialist (in the rows) worked on a given team (in the columns). In short, each column of the  $4793*249$  matrix corresponds to one of the teams of the dependent variable and represents the composition of its members. We used the DENSITY procedure in UCINET VI (Borgatti, Everett, and Freeman, 2002), which allows us to partition the rows of the data matrix into blocks by specifying blocking rules for the density computation. Using the dataset  $4793*249$ , we specified column number 237 (corresponding to production number 489) as a blocking vector. The output is the valued density for the square sub-matrix of size equal to the number of people in team #489, which happens to be 19, thus producing a  $19*19$  matrix, where the  $x_{ij}$  cell denotes the number of times specialists  $i$  and  $j$  worked together in the past. We repeated the blocking procedure 249 times (once for each team),

using and changing the appropriate input dataset according to the specific past time window associated with each team.

### **Two Stage Least Square Analysis (2SLS): Second Stage variables**

**Dependent Variable: Team Performance.** The share of viewers that watched a show is considered the most crucial performance indicator of any TV production. Audience data are collected in Italy by an independent institution (Auditel). Auditel data are used to measure the success or the failure of a TV production. Given the highly skewed nature of audience numbers, we use the natural log of the audience share that watched the TV show as our measure of performance and our dependent variable.

**Independent variables.** We computed Current Structural Holes and Alter Content Homogeneity following the same procedure and measures adopted for the past alters but using the network of current (rather than past) ties to identify the alters of a focal TV production team.

**Controls.** We employ controls of several types in our analysis, beginning with a number of industry-specific factors. First, we control for periodicity effects by using a series of dummies corresponding to the years 1995, 1996, 1997 and 1998, with 1999 being the omitted category. Second, the six TV channels we considered (95 % average of the total audience covered) do not have the same potential for reaching high audience levels. In particular, two major channels broadcast the most significant events and the most popular TV news, and have far higher viewership than the other channels. Accordingly, we include a dummy variable for major channel, which is set to 1 when the production was shown on either major channel and 0 otherwise. In addition, the time slot in which the TV production is broadcast likely affects the potential number of viewers. The highest potential viewership occurs in prime time, which in Italy is 8:00-10:30 p.m. We therefore also include a dummy variable to control for prime time.

In order to control for a potential decay effect of relations in the power of past ties, we also control for the age of all the ties in the network by weighting every tie by the inverse of its age. Specifically, for each team we computed a weighted average of their past relations where the numerator counts the number of ties in each past year and the denominator is the age of the tie (ranging from 1 to 7). Thus, older ties are down-weighted relative to more recent ties. In assessing the sensitivity of this decay function we also evaluated a number of other non-linear functions such as  $\log(x)$ ,  $(x)^{1/2}$ ,  $1/(x)^2$ . The results were consistent across decay functions.

We also included a control for the similarity of the focal production to the industry (*Similarity to Industry*), which captures the extent to which the focal production mirrors the dominant genre. For this control variable, we calculated the average content homogeneity between each focal production and all productions broadcast in the same year as the focal production. Further, in addition to industry and content-related controls, we also control for the size of the team, which is a count of the number of different specialists that comprise it. Finally, we controlled for focal team imitation relative to alters by capturing the average content overlap between the content of each focal team and the content of their current alters (*Conformity to Alters*).

Our final set of controls eliminates the possible effect on performance of certain other characteristics of TV productions. In particular, not all TV productions have similar characteristics: different format exist (e.g. TV movies, soaps, sitcoms) and the number of episodes differ. We control for such task characteristics by computing two additional variables: the number of episodes and a TV movie dummy. The number of episodes variable indicates how many episodes of the TV production were actually broadcast. The TV movie dummy equals 1 when the production is a TV movie and 0 if it is a serial-like production (e.g. a sitcom or a soap).



## RESULTS

Table 1 provides descriptive statistics and correlations for the variables.

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Table 1 about here

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We first tested for the appropriateness of treating Current Structural Holes as an endogenous variable by using the Wu-Hausman F-test (6.14, df (1,214);  $p = 0.01$ ), and Durbin-Wu-Hausman  $\chi^2$  (6.38 (1);  $p = 0.01$ ). Both tests soundly rejected the null hypothesis that Current Structural Holes are exogenous to Performance, indicating that it is appropriate to use a 2SLS specification to address the issue of endogeneity. Moreover, due to the use of several instrumental variables, we checked for potential over-identification in our model. The Sargan statistic provides a measure of instrument relevance, and an inability to reject the null, as in our case, indicates that the model is not overidentified ( $\chi^2 = 6.70$ , (6); ns)

As we reported earlier, because our dependent variable is bounded we used a tobit model with endogenous covariates. We correct for the presence of heteroskedasticity by using the Huber-White sandwich estimator of variance in Stata (Huber, 1967; White, 1980). We also tested for potential autocorrelation, due to the possibility that past structures may be autocorrelated with current structures, using the Durbin-Watson test. We found no evidence of autocorrelation concerns. Further, as reported earlier, we used a two, four and five-year time windows to assess the sensitivity of the models and found generally consistent results.

Results of the tobit 2SLS analysis are reported in Table 2 for both first and second stages. Because our endogenous variable (Current Structural Holes) measured as efficiency is also bounded (our first stage), we ran a series of tobit models for the first stage alone as well (Table 3). For the instrumental variables we hypothesize as causal factors driving the formation of structural holes, the

results of the first stage of tobit 2SLS and a tobit model for the first stage alone are largely consistent. We report and discuss both sets of models below.

Model 1 accounts for controls. Model 2 introduces Past Alter Content Homogeneity. Contrary to our hypothesis H1a, but supportive of H1b, past alter content homogeneity is negatively and significantly associated with current structural holes (Past Alter Content Homogeneity  $\beta = -.003$ ,  $p < .05$  for tobit 2SLS, and  $\beta = -.002$ ,  $p < .01$  for tobit alone) (all hypotheses are tested with coefficient values from the fully specified Model 5). Thus, focal teams with high past alter content homogeneity tend to reduce their structural holes in the current network. Model 3 includes Status derived from past alter performance and Past Team Centrality. The results support the prediction that the prominence of a TV production team, measured as status (H2a) and as past team centrality (H2b), both increase its ability to create structural holes in the current period (Status  $\beta = .011$ ,  $p < .05$  for tobit 2SLS, and  $\beta = .013$ ,  $p < .05$  for tobit alone; Past Team Centrality  $\beta = .001$ ,  $p < .05$  for tobit 2SLS, and  $\beta = .001$ ,  $p < .05$  for tobit alone). Model 4 introduces Past Structural Holes (H3), which we find to be significantly related to Current Structural Holes, ( $\beta = .361$ ,  $p < .01$  for tobit 2SLS, and  $\beta = .389$ ,  $p < .01$  for tobit).

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Table 2 about here

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Model 5 tests the effect of Team Past Cohesion (H4) and shows a significant and negative relationship with the creation of current structural holes ( $\beta = -.043$ ,  $p < .01$  for tobit 2SLS, and  $\beta = -.033$ ,  $p < .01$  for tobit alone). Overall, the first stage accounts for a large proportion of the variance in the formation of current structural holes (pseudo  $R^2 = .41$ ,  $\chi^2 = 176.56$ ,  $p < .01$ ).

In the second stage of the tobit 2SLS, we test competing hypotheses about the positive and negative effects of content homogeneity among alters (H5a and H5b respectively) and of current structural holes (H6) on focal team performance. The idea that the effects of structure and content

are largely independent is supported in our results. Alter content homogeneity enhances performance ( $\beta = 2.189, p < .05$ ), supporting H5a rather than H5b, as do structural holes ( $\beta = 1.69, p < .01$ ), supporting H6.

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Table 3 about here

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## Discussion

Although a vast research stream has examined the outcomes of network structures, relatively little attention has been paid to their origin, particularly at the organizational level of analysis (e.g. Brass et al., 2004). Such an endeavor is important because an understanding of the organizational outcomes of structure and its normative implications is incomplete without also discovering the set of factors and the dynamic processes that gave rise to structure. We focus in particular on the balance between network constraints and the exploitation of opportunities by the focal actor that underlie the creation of favorable network structures. Further, while some research on the creation of organizational ties exists (e.g. Walker et al., 1997; Gulati and Gargiulo, 1999) explanations for the creation of ties do not carry over into an understanding of the creation of structures because they gloss over the nature of the portfolio of ties and importantly, for the presence or absence of ties among the focal actor's alters.

In this paper, we offer and test a theoretical perspective that encompasses opportunity exploitation and structural persistence as underlying drivers of structural holes and their performance outcomes. We show that network actors are presented regularly with opportunities and inducements of varying magnitudes thanks to their positions in the prior social structure. The opportunities created by networks are not just linked concurrently with favorable outcomes at a point in time but project their shadow over the evolution of future networks. Thus, past networks provide actors with experiences, social contexts, and access to knowledge which provide the

opportunities and inducements that may enable actors to enact future structures, while at the same time being constrained by structures from the past. Our deep investigation of the industry context reveals that network structures neither emerge randomly nor as the epiphenomenal outcomes of single, separated dyadic interactions. On the contrary, they are the result of forces which include both the replication of past social interaction by virtue of inertia as well as the exploitation of opportunities provided by past structures.

Further, because our research context comprises temporary networks that are continually being created and dissolved over time we can more clearly disentangle the underlying processes of both prior network-enabled recreation and constraint: the former through the active exploitation of past opportunities, and the latter influenced and limited by the inertia imposed by past structures. As well, since the temporary network organization has been considered as a distinctive characteristic of cultural industries (Starkey, Barnatt and Tempest, 2000; Baker and Faulkner, 1991) our research contributes to better understanding the mechanisms behind their creation and dynamics (Uzzi and Spiro, 2005).

In brief, our results show that alter content homogeneity is not associated with future structural holes, but past status derived from past alter performance, centrality, and past structural holes all lead to the formation of structural holes in future networks. Our two explanations of opportunity exploitation and structural persistence are not necessarily in opposition to each other. We show that actors exploit actively the opportunities related to structural characteristics and content of past networks in enacting the processes that culminate in the creation of future networks and specifically in the achievement of superior network positions for themselves (Nohria, 1992). At same time, by virtue of inertia and constraint, highly embedded structures from the past limit the focal actor's ability to transform past opportunities into valuable current network structures.

In the search for the more specific factors that underlie the genesis of structural holes, we also explicitly tested the assumption inherent in the conceptual underpinnings of the notion of structural holes that nodes' content is reflective of structures which link them (Burt, 2004). To do

so, we examined both the degree to which past alter content homogeneity is associated with structural holes spanned by the focal team in the subsequent network, as well as the independent effects of structural holes and alter content homogeneity on performance. While our prediction with regard to the causal role of content was not supported, and in fact we found the opposite of what we hypothesized, our results confirm that content and structures are related over time, although we find that the connection between them is considerably more complex than that theorized in the literature. In particular, despite the performance benefit of structural holes and the negative effect of content homogeneity on performance, as well as the negative associations between structural holes and content homogeneity, our results supported our alternative hypothesis and indicate that the opportunity created by the past exposure to alters' homogeneity does not provide sufficient inducement to form structural holes in future networks. In particular, past alter content homogeneity reduces rather than increases the propensity for future structural holes, which questions the causal information diversity-seeking rationale implicit in the conventional structural holes logic. As we theorized in our alternative hypothesis, this finding could also have come about because homogeneity among alters may result in fewer structural holes due to the effects of specialization and homophily, such as via the clumping together of genre-driven communities which would lend credence to a structural persistence argument. Future research should more deeply investigate the complex nature of the causal link behind this result.

An important precursor of structural holes we investigated was team status, which we argued presents focal actors with the opportunities to choose amongst potential team members and thereby result in favorable structures. Our results reveal that status derived from past alter performance enhances the propensity to form structural holes. Therefore, rather than a tradeoff, as Podolny (2005) speculated, which would imply a negative relationship, status based on connections of the focal team to high performing teams in the past in fact generate more structural holes in future networks. We theorize that a possible reason for this result may be that the signaling effect of status attracts otherwise disconnected players to the focal actor – in this sense, the network

becomes redundant thanks to the prominence of the focal actor, because alters flock to high status actors independent of the information they may have obtained through network ties. Such market signaling also provides alters with legitimacy and other benefits in the marketplace (Podolny, 1993). In this manner, high status actors are able to enhance their ability to exploit their past positions because of increased power and credibility, resulting in being offered a wider range of choices of alters from which to choose. Our findings with regard to past centrality and its positive effect on structural holes suggests that prominence also confers on the focal actor the power of choice to manage the network in a way that is beneficial in later periods. Alternatively, a more autonomous mechanism may be at work here by which disconnected alters seek out prominent and high status teams.

A further element of our framework points to the role of structural holes in the past which predict the formation of current structural holes. We characterized this as another manifestation of opportunity exploitation by the focal actor. Note that this idea, in our inherently temporary context, implies a teleological reactivation of favorable past structures and therefore is quite different from the notion of structural persistence. This finding means that structural holes spanned by individual specialists in the past give rise to future structural holes for the team of which they are now a part and demonstrates the strength of the focal team's ability to exploit the opportunities that result in favorable social structures.

In addition to opportunity exploitation, our overarching theoretical framework also included constraint arising from structural persistence as contributing to the creation of structural holes. Our finding regarding the effect of past team cohesion suggests a role for the persistence of networks over time because results showed that a major inhibitor of structural holes in the network was the presence of internal cohesion in a previous time period. Lock-in with dense, overlapping ties makes it harder for focal actors to break out of redundant network structures. Thus, individual specialists who have worked together in the past will prefer to be connected in the current period among scattered prior cohesive team members across various current teams. Teams with high team

cohesion in the past will tend to find themselves in tightly-linked structures in subsequent periods because future teams that employ the cohesive members of a prior team will tend to replicate previous connections, resulting in fewer structural holes for the focal team by virtue of structural persistence. Overall, our results provide considerable evidence for the notion that structural entrepreneurs are able to actively exploit opportunities, although inertia and homophily also play a role, in the genesis of network structure.

An important theoretical question that arises in this context is the extent to which the exploitation of opportunities represents active ‘agency’ on the part of the focal actor. White articulates the notion of agency as a mechanism to recreate social structure as “...agency induces additional agency in a chain reaction, emerging a further level of social organization” (1992: 96). Such a notion, as Nohria puts it, “treats actors as purposeful, intentional agents” (1992: 13). Alternatively, networks may emerge more or less autonomously due to actors merely acting on the available choices that have materialized by virtue of their structural positions in prior networks. While we have taken the position in this paper that the exploitation of opportunities created by positions in past networks falls somewhat short of what some have defined as ‘agency behavior’ on the part of network actors (e.g. Emirbayer and Goodwin, 1998), clearly this is an issue that needs theoretical and empirical resolution.

Coming to the performance consequences of structural holes, we found that alter content homogeneity and structural holes both independently improve performance, raising the intriguing possibility that the performance-enhancing effects of structural holes rely on mechanisms other than information diversity. Specifically, control and access benefits may be at work. It is also possible that other results in the literature suggesting that structural holes are not necessarily conducive to innovation via novelty (e.g. Ahuja, 2000) may be the mechanism operating here as well. In the same vein, similarity may also be engendered through common alters rather than direct connections (Burt, 1987). Consequently, rather than being a simple direct relationship between structural holes, diversity, and performance, we uncover a complex pattern of relationships which suggests that

structure and content both independently influence performance, and moreover that it is homogeneity rather than diversity enhances performance.

As we argued before, homogeneity may also enhance outcomes via a more accurate understanding of the skills, and capabilities that other members of the project team might possess, and what has been referred to as knowing who-knows-what, or ‘transactive memory’ in the literature may improve coordination and limit inefficient duplication for the production task at hand (Wegner, 1986). Moreover, high homogeneity can also aid in the team in internalizing common knowledge and enhancing the creative product (Nonaka and Takeuchi, 1995). Another benefit of high homogeneity comes from increased efficiency from knowledge similarity as team members learn from each other, share common codes, and improve on the processes of joint tasks.

Like us, Rodan and Galunic (2004) show independent effects of structural holes and content diversity (alter knowledge heterogeneity) on innovation performance. However, our results differ from Rodan and Galunic in terms of the positive relationship they find between content alters’ diversity and performance, which supports structural holes reasoning, while ours appears to counter it. A possible explanation for this result could be that we investigate the redundancy of the network at the team level of analysis where factors such as efficiency and routines, rather than the heterogeneity of knowledge, may be exerting stronger influences on performance.

An important contribution of our paper is our methodologically clear-cut examination of the effect of structures on performance outcomes that both explores their effects over time and factors in their likely endogeneity. By treating structural holes as explicitly endogenous and using instruments to predict performance, in a longitudinal model, we take a step toward a deeper and more valid understanding of the relationship between network structure and outcomes.

In sum, our efforts in regard to building and testing a theory of the creation of structural holes respond well to the call from Salancik (1995). As he notes:

A network theory that accounts for the appearance and disappearance of structural holes – rather than how they can be used to advantage – and the consequent changes in interactions



over time may provide us with a better understanding how collective action is organized (1995: 346).

Our results help resolve the question of how structural holes emerge in a context of networks in flux over time. We show that opportunity exploitation by the focal organizational actor and structural persistence combine to generate structural holes in future networks, and more generally, we develop and test a holistic theory of the evolution of network structure.

### **Limitations and Directions for Future Research**

Our data do not permit us to map an actor's 'network capability' or its 'relational capability' which may dynamically influence the manner in which opportunity exploitation and structural persistence operate to create and recreate favorable structures. Nevertheless, the fact that even temporary teams we investigate are composed of key individuals who may embody the team's capabilities reduces the extent of this limitation. Moreover, despite the possibility that our results are consistent with actor agency, we do not of course have any direct measures of actor agency, intentions, or motivations. As such, demonstrating agency more directly remains an item for future research. Another limitation of our data is that we only capture group ties as co-memberships among groups and we cannot exclude the possibility that information and knowledge might flow through other kinds of social relationships, including friendships and other personal ties.

Nevertheless, all of the industry experts agreed that co-membership is the only systematic way through which teams are interconnected and in consequence presents an opportunity to systematically collect data over time on co-membership relations for the universe of productions, teams and individual specialists.

Future research should build on our results to further investigate the source of the benefits from spanning structural holes, beyond the access to diverse content, which could also be studied as an outcome of structural holes. In terms of the genesis of network structures, we identified a series of factors that contribute to the creation of structural holes through a longitudinal analysis. Even as

we advance scholarly understanding of network dynamics, much more work is needed to further explore the processes and conditions through which network structures of various kinds, not just structural holes, are formed. A more comprehensive set of explanations may arise from the integration of a structural perspective which we have adopted combined with frameworks used in related fields, such as that on the processes of group and team formation and more broadly, on group dynamics.

Finally, our data are both limited and enhanced by the temporary nature of the teams that we study. The genesis of structural holes in a network with more stable organizations may have somewhat different antecedents, although research on temporary networks provides us the opportunity to better understand the processes and mechanisms behind the creation and evolution of network structures.

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**Table 1: Descriptive Statistics and Correlations**

	Mean	Std. Dev.	1	2	3	4	5	6.	7.	8.	9.
1. Year95	.112	.317									
2. Year96	.169	.375	-.153								
3. Year97	.237	.426	-.205	-.235							
4. Year98	.205	.404	-.183	-.209	-.282						
5. Year99	.277	.448	-.235	-.268	-.362	-.322					
6. Number of Episodes	8.93	25.12	-.053	-.062	-.009	-.061	.149				
7. TV Movie	.305	.461	.055	.074	.032	-.068	-.067	-.197			
8. Prime Time	.867	.340	-.002	.028	-.033	.036	-.021	-.332	.047		
9. Major Channel	.614	.488	.029	.092	-.076	-.066	.037	-.056	-.086	-.008	
10. Team Size	25.36	7.70	-.044	-.217	-.037	.073	.172	.062	-.189	.162	-.18
11. Similarity to Industry	.674	.046	-.149	.372	.019	.145	-.329	-.053	.105	-.052	.13
12. Past Team Centrality	66.17	31.95	.112	.153	.139	-.271	-.097	.015	.256	-.051	-.12
13. Past Structural Holes	.218	.129	-.008	.023	-.010	-.641	.562	.161	-.113	-.091	.10
14. Current Structural Holes	.255	.096	.084	.101	-.323	-.203	.346	.057	.090	.090	.03
15. Status (standardized)	6.47	.10	.054	.067	.092	.687	.117	.080	-.233	-.115	.01
16. Homogeneity among Current Alters	.673	.034	-.260	.568	.108	.141	-.484	-.106	.094	-.013	.10
17. Homogeneity among Past Alters	.678	.005	-.352	-.233	-.054	.405	.126	.024	-.178	.122	.07
18. Conformity to Current Alters	.291	.112	-.129	.268	.225	-.191	-.162	.061	.020	.205	.13
19. Age of Relations	2.68	.617	.524	.106	-.015	-.230	-.238	-.078	.106	.057	.05
20. Past Closure	.245	.450	.012	.100	-.082	.006	-.016	.059	-.137	-.340	.13
21. Team Performance	2.93	.400	.104	-.014	-.068	-.007	.007	-.084	-.130	.191	.05

**Table 1: Descriptive Statistics and Correlations (continued)**

	11	12	13	14	15	16	17	18	19	20	21
11. Similarity to industry											
12. Past Team Centrality	.008										
13. Past Structural Holes	-.210	.258									
14. Current Structural Holes	-.198	.143	.332								
15. Status	.013	.364	.278	.332							
16. Homogeneity among Current Alters	.388	.044	-.270	-.175	-.084						
17. Homogeneity among Past Alters	-.001	.129	-.018	-.303	.025	-.014					
18. Conformity to Current Alters	.210	-.041	.167	.099	.039	.223	.137				
19. Age of Relations	-.064	.053	.111	.133	.024	-.029	-.484	-.067			
20. Past Team Cohesion	.128	-.257	.191	-.128	.256	.142	-.071	.137	.047		
21. Team Performance	-.023	-.131	.092	.166	.087	.072	.014	.053	.180	-.007	

$r > |.105|$ ,  $p < .1$ ;  $r > |.125|$ ,  $p < .05$ ;  $r > |.165|$ ,  $p < .01$

**Table 2: Results of Tobit Model with Endogenous Covariates**

**First Stage**  
**Endogenous**  
**Variable: Current**  
**Structural Holes**

	Model 1	Model 2	Model 3	Model 4	Model 5
Constant	.135 (.181)	.178 (.164)	.165 (.163)	.023 (.163)	.002 (.159)
<b>Controls</b>					
Year95	.109 *** (.033)	.077 ** (.031)	-.012 (.036)	-.017 (.036)	
Year96	.008 (.027)	.003 (.025)	.003 (.025)	.002 (.024)	-.003 (.023)
Year97	-.164 *** (.035)	-.148 *** (.033)	-.149 *** (.033)	-.119 *** (.033)	-.130 *** (.032)
Year98	-.099 *** (.030)	-.046 (.030)	-.056 * (.030)	.017 (.034)	.012 (.033)
Year99	-.088 ** (.042)	-.059 (.039)	-.062 (.040)	-.057 (.038)	-.072 * (.037)
Age of Relations	.034 *** (.009)	.015 (.010)	.021 * (.011)	.020 (.011)	.011 (.010)
Team Size	.003 *** (.001)	.003 *** (.001)	.003 *** (.001)	.005 *** (.001)	.004 *** (.001)
<b>Instrumental Variables</b>					
Homogeneity among Past Alters		-.004 *** (.001)	-.002 ** (.001)	-.001 * (.001)	-.003 ** (.001)
Status			.016 *** (.006)	.012 ** (.006)	.011 ** (.005)
Past Team Centrality			.001 ** (.000)	.001 *** (.000)	.001 ** (.000)
Past Structural Holes				.342 *** (.079)	.361 *** (.077)
Past Team Cohesion					-.043 *** (.011)

**Table 2: Results of Tobit Model with Endogenous Covariates (continued)**

<b>Second Stage</b>					
<i>Dependent Variable: Team Performance</i>					
<b>Controls</b>					
	.304	.941	.953	.880	.866
Constant	(.852)	(.712)	(.707)	(.708)	(.706)
	-.034	-.051	-.052	-.050	-.049
TV Movie	(.056)	(.046)	(.046)	(.047)	(.047)
	-.000	-.000	-.000	-.000	-.000
Episodes	(.001)	(.001)	(.001)	(.001)	(.001)
	.222 ***	.242 ***	.242 ***	.240 ***	.239 ***
Prime Time	(.080)	(.071)	(.071)	(.072)	(.072)
	.495 ***	.492 ***	.491 ***	.492 ***	.492 ***
Major Channel	(.049)	(.044)	(.044)	(.044)	(.044)
Conformity to	.092	-.264	-.270	-.229	-.222
Industry	(.616)	(.529)	(.527)	(.529)	(.529)
Conformity to	-.566 **	-.375 *	-.372 *	-.394 *	-.398 *
Alters	(.264)	(.222)	(.221)	(.221)	(.220)
<b>Independent Variables</b>					
Current Structural	2.760 ***	1.553 ***	1.530 ***	1.670 ***	1.69 ***
Holes	(.788)	(.487)	(.466)	(.437)	(.416)
Homogeneity					
Among Current	2.355 **	2.166 **	2.163 **	2.18 **	2.189 **
Alters	(1.018)	(.906)	(.904)	(.912)	(.913)
$\chi^2$	144.43	175.86	176.92	177.76	179.20
Model Sig.	.000	.000	.000	.000	.000
( <i>p value</i> )					
Log likelihood	222.74	241.62	245.28	254.35	261.36

Standard errors in parentheses.  
 \*  $p < .10$ ; \*\*  $p < .05$ ; \*\*\*  $p < .01$

These models were estimated using the robust variance estimator.

**Table 3: First Stage with Tobit Model**

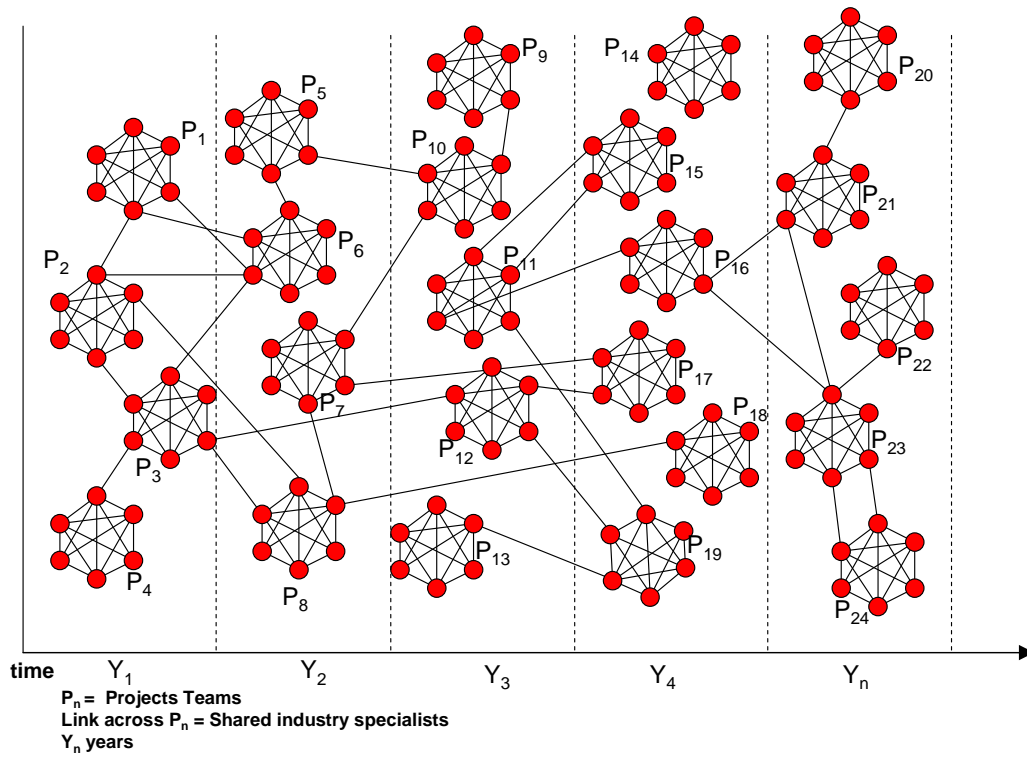
	Model 1	Model 2	Model 3	Model 4	Model 5
Constant	.097 ** (.037)	.183 *** (.037)	.184 *** (.038)	.082 ** (.042)	
<b>Controls</b>					
Year 95	-.020 * (.023)	-.028 (.021)	-.025 (.021)	.006 (.021)	.006 (.021)
Year96	-.008 (.017)	-.023 (.016)	-.019 (.016)	.012 (.016)	.017 (.016)
Year97	-.107 *** (.014)	-.117 *** (.013)	-.112 *** (.013)	-.078 *** (.015)	-.072 *** (.015)
Year98	-.073 *** (.016)	-.036 ** (.016)	-.041 ** (.015)	.052 (.024)	.062 (.024)
Year99	-.070 ** (.033)	-.043 (.022)	-.050 (.014)	-.033 (.025)	-.085 * (.054)
Age of Relations	.029 *** (.010)	.010 (.009)	.017 (.011)	.019 (.010)	.010 (.010)
Team Size	.003 *** (.001)	.003 *** (.001)	.003 *** (.001)	.004 *** (.001)	.004 *** (.001)
<b>Variables</b>					
Homogeneity among Past Alters		-.004 *** (.000)	-.003 ** (.001)	-.001 * (.000)	-.002 *** (.001)
Status			.017 *** (.006)	.012 ** (.006)	.013 ** (.005)
Past Team Centrality			.001 * (.000)	.001 *** (.000)	.001 ** (.000)
Past Structural Holes				.345 *** (.073)	.389 *** (.072)
Past Team Cohesion					-.033 *** (.011)
$\chi^2$	96.68	137.48	145.97	168.02	176.56
Log likelihood	264.78	285.18	289.43	300.45	304.72
Pseudo R <sup>2</sup>	.223	.318	.338	.381	.410
Sig. of $\chi^2$ differences		< .005 ***	< .05 **	< .005 ***	< .005 ***

Standard errors in parentheses.

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

These models were estimated using the robust variance estimator.

**Figure 1: Network Structure of the TV Production Industry**



**Note:** Project Team  $P_{19}$  bridges the past structural hole between  $P_{13}$  and  $P_{12}$ ; Project Team  $P_{23}$  bridges the current structural hole between  $P_{21}$  and  $P_{22}$

## APPENDIX

The operationalization of the TV production content is influenced by the nature of the TV market in Italy which is dominated by so-called ‘generalist’ channels. These are not focused on specific market targets (i.e. age, education, types of productions, and so on) but offer scheduling for mass audiences. Successful TV productions are generally popular movies or series which address both the dominant values and the spirit of the time. In contrast to cinema, TV productions are simpler, more linear in the narration, and generally focused on few popular messages. Thus, we use the following content variables to capture the language, the messages, the narrative and the identity of a TV production. We identified the following twelve variables (see Table 4 below).

**Table 4: Coding of Content**

<b>Variable</b>	<b>Type of variable</b>	<b>Content</b>
<i>Theme</i>	Categorical	Detective; dramatic; life story; friendship; love; family; Bible themed; religion; sport; fantasy; power/money/career; others
<i>Relations</i>	Dummy	1 for love, friendship, kinship, affiliation, affinity, consanguinity, and liaison; 0 otherwise
<i>Values</i>	Dummy	1 for human justice (e.g. story of a crime prosecutors), religion (e.g. stories about the life of saints), freedom and independence, social battles against evil (e.g. citizen or consumers against powerful organizations for environmental protection) or against social prejudice and discrimination; 0 otherwise
<i>Pain</i>	Dummy	1 for stories of disease, suffering, conflicts; 0 otherwise
<i>Power and success</i>	Dummy	1 for power, money, career, social elites; 0 otherwise
<i>Profession of characters</i>	Categorical	Dominant professions in which the story is set
<i>Positive or negative characters</i>	Ordinal	Weighted number of protagonists, antagonists, secondary protagonists
<i>Ending</i>	Categorical	The nature of the epilogue: happy, ambiguous, unhappy
<i>Setting</i>	Categorical	Context in which the story has been located: Italy, Europe, abroad
<i>Time period</i>	Categorical	Time period in which the story is set
<i>Schema</i>	Categorical	The conflict schema in the sentimental relationships (Holsti, 1969)