



HOW DO PERSONAL RESEARCH-**NETWORKS INFLUENCE INNOVATION?** THE BIOMEDICAL CONTEXT

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MOTIVATION OF THIS RESEARCH

Motivations

- To investigate network configurations that are most conducive to knowledge creation
 - Multiple types of network configurations: e.g. structure and composition
 - Multiple types of knowledge outcomes: e.g. scientific and technological
 - ... mixed findings and unsolved conceptual puzzles
- Networks in the context of biomedical research.
 - Translational Research has become a high policy priority with the aim to improve healthcare by strengthening research collaborations between basic and clinical scientists
 - ... but there is a lack of consensus about whether and to what extent current initiatives to support Translational Res. have been really effective
- Improve our understanding of how (biomedical) research networks work



Social network literature

- "People who do better are somehow better connected" (Burt, 2000)
 - Holding a particular position in a network can be an asset in its own right, as it influences the amount of resources and opportunities available
- However, it is not clear what <u>better connected</u> means:
 - being <u>better</u> connected does not necessarily mean being <u>more</u> connected
 - There are different mechanisms to reach advantageous positions in a network
- Two critical aspects of network configurations
 - Structure: Dense (Coleman, 1988) vs. Sparse (Burt, 1992; Granovetter, 1973) networks
 - Composition: Homogeneous vs. Heterogeneous actors (Fleming et al., 2007; Reagans & McEvily, 2003)



Network structure: Dense vs. Sparse networks

Dense Networks



Networks where everyone is highly connected to each other

- Fast access to information
- Reliable communication, people trust each other (cheating and non-reciprocity are socially sanctioned)

Sparse Networks



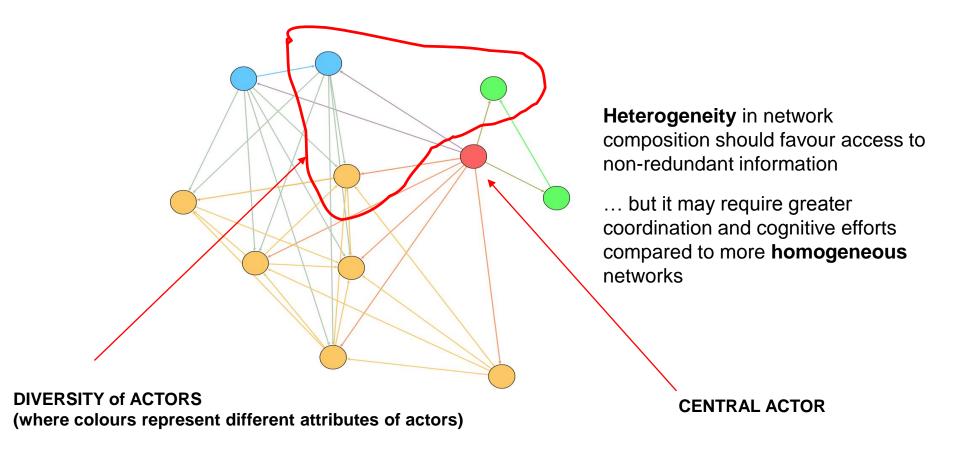
Few connections between alters / More opportunities to act as a bridge between actors - brokers - and control information flows

- High access to non-redundant information
- Unique conditions to identify new opportunities

Sparse networks should increase the exposure to different approaches and sources of information, but they may involve lack of mutual trust and slower circulation of information among partners compared to **dense** networks.

Network composition: homogeneous vs. heterogeneous actors

Network composition refers to the diversity of actors involved in a personal network





Personal network <u>structure</u> / <u>composition</u> and <u>medical innovation</u>

- Actors who have access to diverse sources of information and knowledge as a result of:
 - holding brokerage positions connecting actors who otherwise would be disconnected
 - building ties to heterogeneous actors
 - ... are expected to have an advantage for knowledge creation (Burt,1992; Fleming et.al, 2007; Reagans & McKevily, 2003)
- However, actors may face increasing difficulties to benefit from sparse or heterogeneous networks due to:
 - potential lack of mutual trust and weakened expectations on the credibility of partners
 - potential lack of shared cognitive frames and risks of misperceptions
- ... We expect <u>sparse</u> networks / <u>heterogeneity</u> in network composition to facilitate medical innovation up to a point, beyond which enlarging the range of <u>disconnected</u> / <u>heterogeneous</u> relationships can be either ineffective or detrimental for innovation (Baer, 2010; Fang et al., 2010; McFadyen & Cannella, 2004; ter Wal et al., 2013).

Hypothesis 1: Scientists with personal networks characterized by a high degree of **brokerage** will be more likely to engage in **medical innovation**. Engagement in innovation will be maximized at intermediate levels of brokerage (inverted U-shape).

Hypothesis 2: Scientists with personal networks characterized by high degree of actor **heterogeneity** will be more likely to engage in **medical innovation**. Engagement in innovation will be maximized at intermediate levels of actor diversity (inverted U-shape).

Social network research and <u>characteristics of actors in the network</u>

 Network research has often treated actors as undifferentiated: e.g. cognitive hollow (Phelps et al., 2012)

However:

- Differences in Individual behaviour cannot be solely explained by structure-level characteristics. We need to bring the individual back when conducting social network research (Ibarra, Kilduff & Tsai, 2005)
- Individual differences might refer to:
 - Cognitive frames and skills (Rotolo & Messeni-Petruzzelli, 2012)
 - Personality traits (Fini et al., 2012)
 - Motivations and attitudes (Mehra et al 2001)



Social network research and <u>characteristics of actors in the network</u>

We consider two types of characteristics regarding central actors:

Cognitive skills

Engagement in medical innovation requires that scientists should be familiar with a combination of basic and clinical skills (Hobin et al, 2012).

Hypothesis 3: Breadth of cognitive skills will have a positive relationship with the scientists' degree of engagement in medical innovation.

Perceived impact on beneficiaries

Perceived impact on beneficiaries is the degree to which individuals are aware that their own actions have the potential to improve the welfare of others (Grant, 2007, 2008). This awareness is claimed to exert an influence on individuals' disposition to channel this perception into outcomes.

Hypothesis 4: The perceived impact of research on patients and medical practitioners will have a positive relationship with the scientists' degree of engagement in medical innovation.



RESEARCH CONTEXT AND METHODS

Spanish Biomedical Research Networking Centers (CIBERs) are formal network platforms created by the Spanish Ministry of Health in 2007.

Aims of the CIBER networks:

- Foster research collaboration by bringing together research groups from universities, hospitals, research centres and firms working on similar pathologies.
- Organize biomedical research around nine broad range of pathologies of critical interest for the Spanish National Health System:
 - Bioengineering, Biomaterials and Nanomedicine
 - Diabetes and Metabolic Associated Diseases
 - Epidemiology and Public Health
 - Hepatic diseases
 - Mental Health
 - Neurodegenerative diseases
 - Obesity and Nutrition
 - Rare Diseases
 - Respiratory Diseases



RESEARCH CONTEXT AND METHODS (II)

SURVEY DATA

- Sample frame for the study:
 - All biomedical scientists and technicians belonging to research groups in each of the nine CIBER networks (4,758 individuals)

Implementation of a survey

- We designed a questionnaire to collect information on the following aspects
 - collaborative network (external to the scientist' research team)
 - individual attributes of scientists
 - degree of engagement in multiple activities related to medical innovation
- Using email addresses, scientists were invited to participate an on-line survey (between April and June, 2013)
- Overall response rate = 27.5 % (1,309 valid responses)
 - Non-response bias tests by type of institution, group size, status and CIBERs

SECONDARY SOURCES

Records of patent applications from Pls (period 1990-2010)



VARIABLES

Dependent variable: Degree of engagement in (multiple types of) medical innovation

We asked respondents to report "how many times" they have participated in any of the following activities during the year 2012.

Included items in the questionnaire

Patent applications for new drugs

Licenses from patents

Participation in spin-off

Clinical trials phases I, II or III for new drugs development

Clinical trials phase IV for new drugs development

Clinical trials phase IV for new diagnostic techniques

Clinical guidelines for healthcare professionals

Clinical guidelines for patients

Patent applications for new diagnostic techniques

Clinical trials phases I, II or III for new diagnostic techniques

Clinical guidelines for the general population (prevention)



VARIABLES

Dependent variable: Degree of engagement in (multiple types of) medical innovation

We asked respondents to report "how many times" they have participated in any of the following activities during the year 2012.

Included items in the questionnaire	Categories
Patent applications for new drugs	
Licenses from patents	Invention and
Participation in spin-off	Commercialization
Clinical trials phases I, II or III for new drugs development	
Clinical trials phase IV for new drugs development	New Drug Development
Clinical trials phase IV for new diagnostic techniques	
Clinical guidelines for healthcare professionals	
Clinical guidelines for patients	Clinical Guidelines
Patent applications for new diagnostic techniques	
Clinical trials phases I, II or III for new diagnostic techniques	Diagnostics and Prevention
Clinical guidelines for the general population (prevention)	



• DV: Degree of engagement in Med. Innov.- ranges between 0 and 3 according to the participation in the four types of medical innovation:

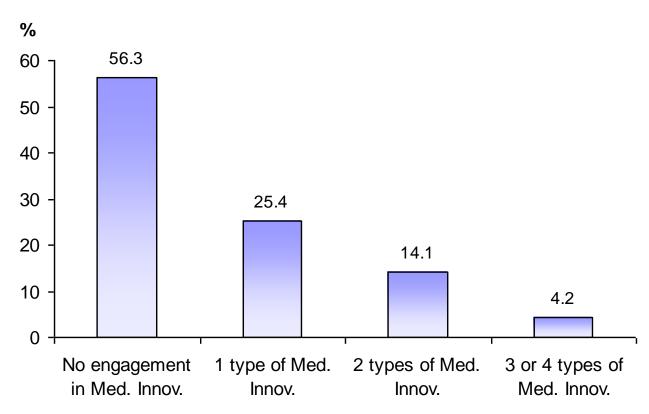
"0": No participation in any of the four types of innovation

"1": Participated at least once in one of the four types of innovation

"2": Participated at least once in two of the four types of innovation

"3": Participated at least once in three or four innovation types.

Distribution of respondents across the categories of the DV (%)





Independent variable I: Ego-network brokerage

We measured network brokerage as the rate of actual connections / potential connections between each respondents' contacts from outside her research group.

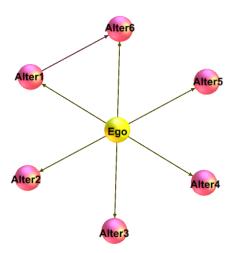
Personal (ego) network brokerage as: 1 - Numb

Number of alter-alter ties
total number of possible
alter-alter ties

Min = 0 (lowest brokerage) Max = 1 (highest brokerage)

Scientist A

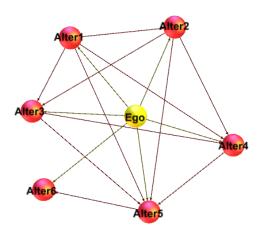
6 alters reported Brokerage score = 0,933



High Brokerage (sparse network)

Scientist B

6 alters reported Brokerage score = 0,267

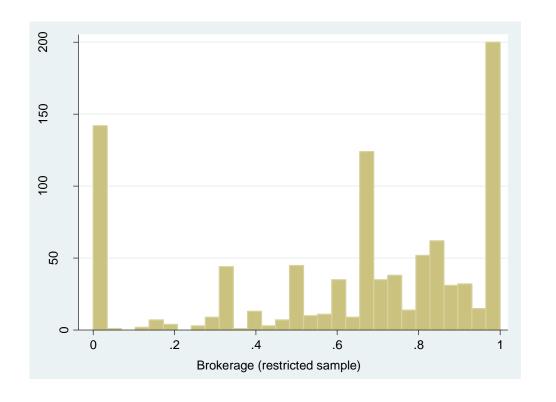


Low Brokerage (dense network)



Independent variable I: Ego-network brokerage

Frequency of scientists according to their brokerage score



Mean: 0.63 Median: 0.70 Mode: 1.00 Min: 0.00 Max: 1.00 (figures for actors who report 2 or more

external alters)

Frequencies are largest at the extremes of the distribution: scores of 0 and 1



MEASURE of Network Composition

Independent variable II: Network_Range

This measure is built from another question in the survey, asking for the following information about the contacts cited by the respondent:

"Indicate the sector or professional field of the persons you have cited as being a particularly important source of information or advice for your research activities" (drop-down menu)

	Basic Scientist (NHS, Uni)	Clinical Scientist (NHS, Uni)	Medical Doctor (not involved in research)	Patient or Patient Associations	Industry / Private Sector	Public Administ	Others (specify)
Alter 1							
Alter 2							
Alter 3							
Alter 4							
Alter 5							
Alter 6							
Alter 7							
Alter 8							
Alter 9							
Alter 10							



MEASURE of Network Composition

Independent variable II: Network_Range

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	Basic Scientist (NHS, Uni)	Clinical Scientist (NHS, Uni)	Medical Doctor (not involved in research)	Patient or Patient Associations	Industry / Private Sector	Public Administ	Others (specify)
Alter 1							
Alter 2							
Alter 3							
Alter 4							
Alter 5							
Alter 6							
Alter 7							
Alter 8							
Alter 9							
Alter 10							

We grouped the alters in these 4 categories.

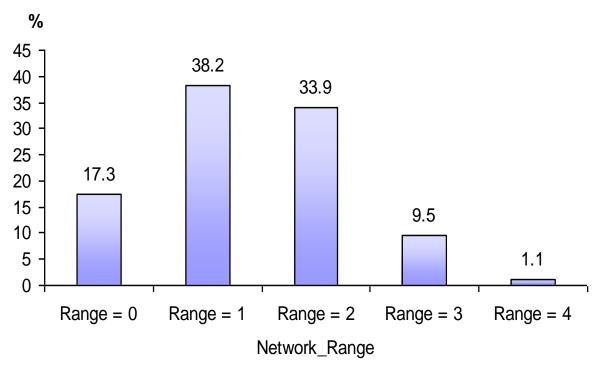


MEASURE of Network Composition

Independent variable II: Network_Range

We constructed our variable Network_Range as the count of sector / professional categories of the alters that compose an individual's network.

Network_range takes values from 0 (no external contacts) to 4 (external contacts belonging to the four categories of sectors or professional activity).



About 70% of our respondents report having external contacts who belong to 1 or 2 distinct categories of sectors or professional activity.

MEASURE of Cognitive Skills

Independent variable III: Breadth of cognitive skills

The survey included the following question:

"Have you received, through your career, specific training in one or more of the following activities?" (tick where appropriate)

Design of clinical trials	
Design of clinical guidelines	
State-of-the-technology in your field of research	
Clinical pharmacology	
Biostatistics	
Molecular biology	
Experimental methods	
Experimentation with animals	
Studies with control groups	

Cognitive Breadth: Measured as the count of areas of 'specific training'

Median: 3.00 Mode: 2.00 Min: 0.00 Max: 9.00

Mean: 2.71



MEASURE of Perceived Impact on Beneficiaries

Independent variable IV: Perceived impact on beneficiaries

The survey included the following question:

"Please, indicate the extent to which you consider that the following collectives benefit more directly from the results obtained from your research activities" (responses according to a 7 point Likert Scale – from 'not at all' to 'very much')

Collectives	1	2	3	4	5	6	7
Patients							
Clinical Practitioners							
Patients' relatives							

- We averaged the responses to the three items to create a composite indicator of the perceived clinical impact of the research activities (Cronbach's Alpha = 0, 78)

Perceived impact on beneficiaries: Mean: 4.44 / Median: 4.50 / Mode: 5.00 / Min: 1.00 / Max: 7.00



Control Variables and Econometric Methods

Control variables

Individual level:

- Age & Gender
- PhD degree
- Size of external network
- Patent applications over period 1990-2010

Organizational and Institutional

- Size of the research team
- Institutional affiliation: University, Hospital, PROs and Others
- Type of CIBER

Econometric Method

Ordered Probit / Logit, Fractional Logit and OLS regression methods

- Dependent variable that ranges between 0 and 3 (Ordered Logit / Probit)
- Re-scale the variable to obtain a measure between 0 and 1: $P_i = (Y_i Y_{min}) / (Y_{max} Y_{min})$ (F.Log.)

Consider two samples

- Complete sample, controlling for cases with zero or one external contact (1111 obs.)
- Restricted sample: considering only those cases who report having 2 or more external contacts (820 obs.)

Dependent variable: Degree of engagement in medical innovation activities (outcome values: 0 - 3)

T	otal Sample	(1111 obs.)	Restricted Sample (820 obs.)*				
(1a)	(2a)	(3a)	(4a)	(1b)	(2b)	(3b)	(4b)

Predictor Variables

Ego Net. Brokerage

Ego Net. Brokerage²

Network Range

Network Range²

Cognitive Breadth

Perc.Impact Benef.

Control Variables

Age

PhD

Large Ego-Network

Past Patent Applicat.

Gender (female=1)

Group Size

University

Hospital

PROs

CIBER (8 dummies)

Ext_net.< 2 (dummy)

Ps-R² (Cragg-Uhler)

^{*} p < 0.1, ** p < 0.05, *** p < 0.01 / * N. Observations = 820 (scientists who reported at least 2 external contacts)



	Total Sample (1111 obs.)			Restricted Sample (820 obs.)*				
•	(1a)	(2a)	(3a)	(4a)	(1b)	(2b)	(3b)	(4b)
Predictor Variables								
Ego Net. Brokerage								
Ego Net. Brokerage ²								
Network Range								
Network Range ²								
Cognitive Breadth								
Perc.Impact Benef.								
Control Variables								
Age	0.021***				0.022***			
PhD	0.092				-0.001			
Large Ego-Network	0.195**				0.194**			
Past Patent Applicat.	0.037***				0.033*			
Gender (female=1)	-0.361***				-0.374***			
Group Size	0.004				0.005			
University	0.003				-0.098			
Hospital	0.805***				0.766***			
PROs	0.133				0.112			
CIBER (8 dummies)	Included				Included			
Ext_net.< 2 (dummy)	-0.143							
Ps-R ² (Cragg-Uhler)	0.24				0.22			•
* = .0.4 ** = .0.05 **	* n . 0 04 / * N	I Observation	000 (00)	tioto who won	owtool of locat O	ovetownol ook	ta ata)	OCOL

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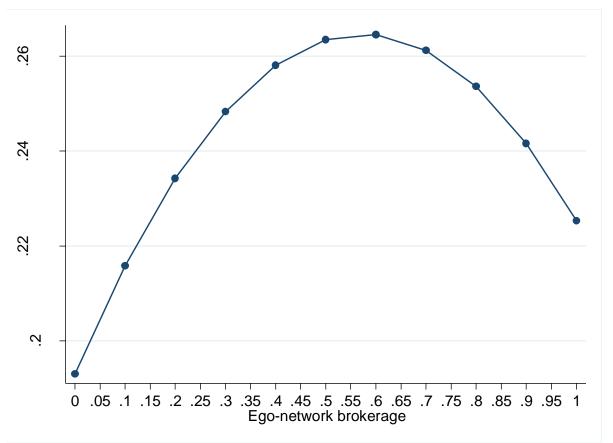
	T	otal Sample	(1111 obs.)		Restricted Sample (820 obs.)*			
•	(1a)	(2a)	(3a)	(4a)	(1b)	(2b)	(3b)	(4b)
Predictor Variables								
Ego Net. Brokerage		1.103**				1.144**		
Ego Net. Brokerage ²		-0.917**				-0.937**		├ H1
Network Range								
Network Range ²								
Cognitive Breadth		0078***				0.091***		
Perc.Impact Benef.		0.188***				0.207***		
Control Variables								
Age	0.021***	0.021***			0.022***	0.021***		
PhD	0.092	0.092			-0.001	-0.013		
Large Ego-Network	0.195**	0.025			0.194**	0.013		
Past Patent Applicat.	0.037***	0.044***			0.033*	0.036**		
Gender (female=1)	-0.361***	-0.376***			-0.374***	-0.394***		
Group Size	0.004	0.006			0.005	0.007		
University	0.003	0.053			-0.098	-0.042		
Hospital	0.805***	0.736***			0.766***	0.701***		
PROs	0.133	0.157			0.112	0.153		
CIBER (8 dummies)	Included	Included			Included	Included		
Ext_net.< 2 (dummy)	-0.143	0.081						
Ps-R ² (Cragg-Uhler)	0.24	0.30			0.22	0.29		_

 $^{^*}$ p < 0.1, ** p < 0.05, *** p < 0.01 / * N. Observations = 820 (scientists who reported at least 2 external contacts)



RESULTS: NETWORK BROKERAGE AND MEDICAL INNOVATION

Curvilinear relationship between network brokerage and engagement in medical innovation



The highest participation in medical innovation happens at intermediate levels of network brokerage



	Total Sample (1111 obs.)				Res	tricted San	nple (820 ol	os.)*
•	(1a)	(2a)	(3a)	(4a)	(1b)	(2b)	(3b)	(4b)
Predictor Variables								
Ego Net. Brokerage		1.103**				1.144**		
Ego Net. Brokerage ²		-0.917**				-0.937**		
Network Range			0.164**				0.133	-
Network Range ²			-0.027				-0.010	
Cognitive Breadth		0078***	0.075***			0.091***	0.087***	
Perc.Impact Benef.		0.188***	0.181***			0.207***	0.195***	
Control Variables								
Age	0.021***	0.021***	0.021***		0.022***	0.021***	0.021***	
PhD	0.092	0.092	0.084		-0.001	-0.013	-0.019	
Large Ego-Network	0.195**	0.025	0.041		0.194**	0.013	0.040	
Past Patent Applicat.	0.037***	0.044***	0.044***		0.033*	0.036**	0.036*	
Gender (female=1)	-0.361***	-0.376***	-0.386***		-0.374***	-0.394***	-0.404***	
Group Size	0.004	0.006	0.006		0.005	0.007	0.006	
University	0.003	0.053	0.067		-0.098	-0.042	-0.028	
Hospital	0.805***	0.736***	0.738***		0.766***	0.701***	0.702***	
PROs	0.133	0.157	0.176		0.112	0.153	0.166	
CIBER (8 dummies)	Included	Included	Included		Included	Included	Included	
Ext_net.< 2 (dummy)	-0.143	0.081	0.084					
Ps-R ² (Cragg-Uhler)	0.24	0.30	0.30		0.22	0.29	0.29	_

 $^{^{*}}$ p < 0.1, ** p < 0.05, *** p < 0.01 / * N. Observations = 820 (scientists who reported at least 2 external contacts)



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Ego Net. Brokerage ²		-0.917**				-0.937**			
Network Range			0.164**				0.133		
Network Range ²			-0.027				-0.010		
Cognitive Breadth		0078***	0.075***			0.091***	0.087***	Н3	
Perc.Impact Benef.		0.188***	0.181***			0.207***	0.195***	H4	
Control Variables								•	
Age	0.021***	0.021***	0.021***		0.022***	0.021***	0.021***		
PhD	0.092	0.092	0.084		-0.001	-0.013	-0.019		
Large Ego-Network	0.195**	0.025	0.041		0.194**	0.013	0.040		
Past Patent Applicat.	0.037***	0.044***	0.044***		0.033*	0.036**	0.036*		
Gender (female=1)	-0.361***	-0.376***	-0.386***		-0.374***	-0.394***	-0.404***		
Group Size	0.004	0.006	0.006		0.005	0.007	0.006		
University	0.003	0.053	0.067		-0.098	-0.042	-0.028		
Hospital	0.805***	0.736***	0.738***		0.766***	0.701***	0.702***		
PROs	0.133	0.157	0.176		0.112	0.153	0.166		
CIBER (8 dummies)	Included	Included	Included		Included	Included	Included		
Ext_net.< 2 (dummy)	-0.143	0.081	0.084						
Ps-R ² (Cragg-Uhler)	0.24	0.30	0.30		0.22	0.29	0.29		

^{*} p < 0.1, ** p < 0.05, *** p < 0.01 / * N. Observations = 820 (scientists who reported at least 2 external contacts)



-	T	otal Sample	(1111 obs.)		Rest	tricted Sam	ple (820 o	bs.)*
-	(1a)	(2a)	(3a)	(4a)	(1b)	(2b)	(3b)	(4b)
Predictor Variables								
Ego Net. Brokerage		1.103**		0.979**		1.144**		1.048**
Ego Net. Brokerage ²		-0.917**		-0.386*		-0.937**		-0.875**
Network Range			0.164**	0.125**			0.133	0.097
Network Range ²			-0.027				-0.010	
Cognitive Breadth		0078***	0.075***	0.075***		0.091***	0.087***	0.088***
Perc.Impact Benef.		0.188***	0.181***	0.183***		0.207***	0.195***	0.201***
Control Variables								
Age	0.021***	0.021***	0.021***	0.021***	0.022***	0.021***	0.021***	0.021***
PhD	0.092	0.092	0.084	0.086	-0.001	-0.013	-0.019	-0.018
Large Ego-Network	0.195**	0.025	0.041	-0.035	0.194**	0.013	0.040	-0.032
Past Patent Applicat.	0.037***	0.044***	0.044***	0.044***	0.033*	0.036**	0.036*	0.036**
Gender (female=1)	-0.361***	-0.376***	-0.386***	-0.387***	-0.374***	-0.394***	-0.404***	-0.405***
Group Size	0.004	0.006	0.006	0.006	0.005	0.007	0.006	0.007
University	0.003	0.053	0.067	0.060	-0.098	-0.042	-0.028	-0.037
Hospital	0.805***	0.736***	0.738***	0.733***	0.766***	0.701***	0.702***	0.697***
PROs	0.133	0.157	0.176	0.167	0.112	0.153	0.166	0.159
CIBER (8 dummies)	Included	Included	Included	Included	Included	Included	Included	Included
Ext_net.< 2 (dummy)	-0.143	0.081	0.084	0.184				
Ps-R ² (Cragg-Uhler)	0.24	0.30	0.30	0.30	0.22	0.29	0.29	0.29

^{*} p < 0.1, ** p < 0.05, *** p < 0.01 / * N. Observations = 820 (scientists who reported at least 2 external contacts)



PRELIMINARY CONCLUSIONS

- What type of personal networks are most conducive to innovation?
- Our results suggest that:
 - A. The structure of scientists' collaboration network <u>does</u> influence innovation
 - ... but it is important to keep an **appropriate balance** between sparse and dense network structures
 - Scientists devoting efforts to cultivate a sparse network are more strongly engaged in medical innovation
 - However, maintaining sparse networks may undermine trust or involve coordination difficulties
 - Most effective network structures combine elements associated to both dense and sparse networks



PRELIMINARY CONCLUSIONS

B. We find partial evidence of a positive impact of Network range on innovation

- Networks composed of heterogeneous actors seem to be conducive to innovation among scientists
 - ✓ However, contrary to ego-network brokerage, the potential benefits of having a divers network does not show decreasing returns.
 - ✓ These results do not hold for the restricted sample: low variability of network range (85% of obs. have a range score of 1 or 2).

C. Network structures should be analyzed in conjunction with Individual attributes:

- Cognitive breadth: the higher the diversity of (basic & clinical) skills, the higher the probability of scientists to engage in medical innovation
 - More Inter-disciplinary univ. degree programs bridging basic and clinical research requires sets of skills that are not typically offered by traditional curricula
- Perceived impact on beneficiaries: scientists who are particularly aware of the positive impact they exert on <u>patients</u> and <u>clinical practitioners</u> are more prone to engage in multiple forms of medical innovation
 - ✓ Our results support Soc. Psych. Lit. suggesting that when individuals perceive that their actions have an impact on beneficiaries, they are particularly motivated to make a positive difference in the wellbeing of these beneficiaries (e.g. developing new med. treatments)







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VARIABLES (II)

Independent variable I: Ego-network brokerage

"Write down the names of those persons (up to ten) from outside your research group that are particularly important for the advancement of your research activities"

Ego-network brokerage

Number of alter-alter ties total number of-possible alter-alter ties

Min = 0 (lowest brokerage) Max = 1 (highest brokerage)

Independent variable II: Breadth of cognitive skills

"Have you received, though your career, training on one or more of the following activities?"

Battery of 8 skills. E.g.: "development of clinical trials", "biostatistics", "molecular biology", "experimental methods"

Independent variable III: Perceived impact on beneficiaries

"Please, indicate the extent to which the following collectivities benefit more directly from the results obtained from your research activities" (Likert scale, 1 -7)

- a) Research community,
- b) Patients;
- c) Clinical practitioners
- d) Vulnerable social groups



RESEARCH CONTEXT AND METHODS (II)

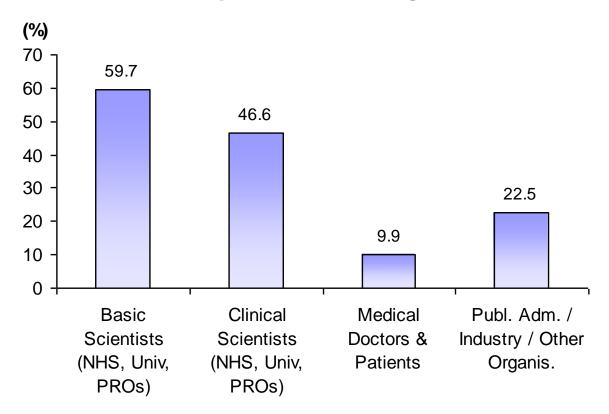
Response rates by CIBER:

CIBER	Population surveyed	Nº Complete Returned Questionnaires	Response rates (%)
BBN-Bioeng.,Biomaterials & Nanomed.	872	238	27.3
DEM -Diabetes & Metabolic A. Diseases	331	96	29.0
EHD-Hepatic Diseases	459	154	33.6
ER-Rare Diseases	517	177	34.2
ES-Respiratory Diseases	439	159	36.2
ESP-Epidemiology & Public Health	610	107	17.5
NED -Neurodegenerative Diseases	750	186	24.8
OBN-Obesity & Nutrition	303	71	23.4
SAM-Mental Health	477	121	25.4
Total	4758	1309	27.5



Independent variable II: Network_Range

Proportion of individuals who report having <u>at least one</u> contact corresponding to each of the four sector / professional categories



E.g.: About 60% of our respondents report that at least one of their (external) informants were Basic Scientists.

.... Should "optimal" network configurations lay somewhere in-between?

Personal networks where actors enjoy the advantages of both types of structures





MEASURE of Network Structure

Independent variable I: Ego-network brokerage

The survey included the following question:

"Write down the names of those persons (up to ten) from outside your research group who have been a particularly important source of information or advice for the advancement of your research activities in 2012"

A subsequent question was then activated with the following matrix (size depending on the number of alters reported) asking for the following information:

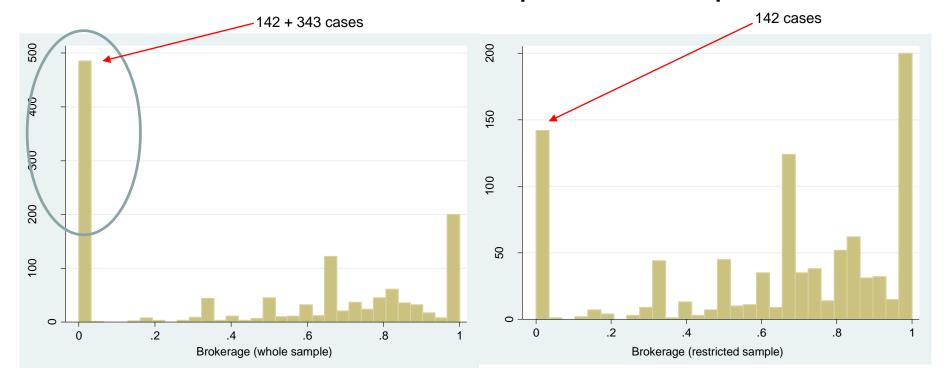
"Indicate if, according to your knowledge, the persons you have cited exchange information or advice with each other, in connection with their professional activities" (tick as many as appropriate)

	Alter 1	Alter 2	Alter 3	Alter 4	Alter 5	Alter 6	Alter 7	Alter 8	Alter 9
Alter 2									
Alter 3									
Alter 4									
Alter 5									
Alter 6									
Alter 7									
Alter 8									
Alter 9									
Alter 10									



Independent variable I: Ego-network brokerage

Network structure measures are computed for two samples



Whole sample: N. Obs. 1309

• Includes all obs., including those cases reporting zero or 1 external contacts

Restricted sample: N. Obs. 949

Includes only those cases reporting2 or more external contacts

PRELIMINARY CONCLUSIONS

Avenues for further research:

- Variety of indicators for medical innovation: (i) drug development, (ii) clinical guidelines, (iii) invention and commercialization: and (iv) diagnostics/prevention:
 - Results by type of medical innovation
 - Distinct explanatory factors from different types of brokerage
- Moderating factors:
 - ✓ Interplay between Structure and Composition
 - ✓ Network configuration Individual Attributes
- Scientific performance:
 - ✓ to explore whether scientific excellence is a predictor of engagement in med. Innov.
 - to examine whether scientific performance could contribute to enact personal networks
- Differences in network configurations for <u>innovation</u> and <u>scientific discoveries</u>



PRELIMINARY CONCLUSIONS

- Individual attributes should be explicitly considered as they critically contribute to knowledge creation (in addition to network features):
 - Cognitive breadth: the higher the diversity of (basic & clinical) skills, the higher the probability of scientists to engage in medical innovation
 - More Inter-disciplinary univ. degree programs bridging basic and clinical research requires sets of skills that are not typically offered by traditional curricula
 - Perceived impact on beneficiaries: scientists who are particularly aware of the positive impact they exert on <u>patients</u> and <u>clinical practitioners</u> exhibit a stronger engagement in multiple forms of medical innovation
 - ✓ Our results support Soc. Psych. Lit. suggesting that when individuals perceive that their actions have an impact on beneficiaries, they become particularly motivated to make a positive difference in the wellbeing of these beneficiaries (developing new med. treatments)
 - ✓ Implementation of mechanisms to increase scientists' awareness of the practical impact on patients and clinical practitioners, to foster their participation in medical innovation activities: particularly among <u>basic</u> scientists



-	Total Sample (1111 obs.))	Restricted Sample (820 obs.)*			bs.)*
•	(1a)	(2a)	(3a)	(4a)	(1b)	(2b)	(3b)	(4b)
Predictor Variables					1			
Ego Net. Brokerage		1.103**		0.979**		1.144**		1.048**
Ego Net. Brokerage ²		-0.917**		-0.386*		-0.937**		-0.875**
Network Range			0.164**	0.125**			0.133	0.097
Network Range ²			-0.027				-0.010	
Cognitive Breadth		0078***	0.075***	0.075***		0.091***	0.087***	0.088***
Perc.Impact Benef.		0.188***	0.181***	0.183***		0.207***	0.195***	0.201***
Control Variables]			
Age	0.021***	0.021***	0.021***	0.021***	0.022***	0.021***	0.021***	0.021***
PhD	0.092	0.092	0.084	0.086	-0.001	-0.013	-0.019	-0.018
Large Ego-Network	0.195**	0.025	0.041	-0.035	0.194**	0.013	0.040	-0.032
Past Patent Applicat.	0.037***	0.044***	0.044***	0.044***	0.033*	0.036**	0.036*	0.036**
Gender (female=1)	-0.361***	-0.376***	-0.386***	-0.387***	-0.374***	-0.394***	-0.404***	-0.405***
Group Size	0.004	0.006	0.006	0.006	0.005	0.007	0.006	0.007
University	0.003	0.053	0.067	0.060	-0.098	-0.042	-0.028	-0.037
Hospital	0.805***	0.736***	0.738***	0.733***	0.766***	0.701***	0.702***	0.697***
PROs	0.133	0.157	0.176	0.167	0.112	0.153	0.166	0.159
CIBER (8 dummies)	Included	Included	Included	Included	Included	Included	Included	Included
Ext_net.< 2 (dummy)	-0.143	0.081	0.084	0.184				
Ps-R ² (Cragg-Uhler)	0.24	0.30	0.30	0.30	0.22	0.29	0.29	0.29

 $^{^*}$ p < 0.1, ** p < 0.05, *** p < 0.01 / * N. Observations = 820 (scientists who reported at least 2 external contacts)



VARIABLES

• Proportion of scientists who participate in the different types of medical innovation, by type of Institution (%)

	Invention & Commercializ.	Drug Development	Clinical Guidelines	Diagnostics & Prevention	Total obs.
Universisty	19,2	7,5	11,7	8,8	386
Hospitals	12,0	41,4	47,8	12,5	409
Public Research Centres	15,5	8,8	9,4	10,3	341
Private Research Centres & Others	15,2	8,8	12,0	7,2	125
Total	15,5	19,0	22,8	10,2	1261

