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Does co-authorship matter for scientific productivity? Evidence from geography's top journals

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DOES CO-AUTHORSHIP MATTER FOR SCIENTIFIC PRODUCTIVITY? EVIDENCE FROM GEOGRAPHY'S TOP JOURNALS

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Abstract

The aim of this paper is to investigate the relationship between scientific productivity and collaborative behaviours (formal and informal). Despite the different approaches available in the literature, we will focus on what we call the “relational trend”: our goal is to detect some of the factors which might affect researchers’ productivity, considering “relational variables”. In particular, the tendency to write papers in co-authorship will be used as a proxy of formal scientific collaboration amongst scientists, while the number of acknowledgments will be assumed as a proxy of the scientist’s ability to build informal collaboration networks. Both co-authorships and acknowledgments indices are interpreted as two of the main forces which could affect and drive scientific production, apart from individual talent. Using the dataset developed by Togni (2009) which collects data about geographers’ publications on the Top Journals in the years 2000-2007, an econometric analysis using two-stage least squares has been performed, in order to regress *productivity* on a series of other indices, including (amongst the others) a typical SNA index of centrality (*betweenness centrality*). Three results clearly emerged from the analysis: co-authorships networks affect productivity in a negative way, but a variety of co-authors may increase geographers’ productivity; on the contrary, informal influence (acknowledgments network) on productivity seems not to have any effect on productivity. Finally, a trade-off between the transaction costs from the collaboration and the need to mutual exchange of skills and knowledge complementarities which boost scientists to vary their co-authors.

Keywords: SCIENTIFIC NETWORKS, CO-AUTHORSHIPS, ACKNOWLEDGMENTS, SCIENTIFIC PRODUCTIVITY, TOP JOURNALS, GEOGRAPHERS, NETWORK ANALYSIS, TWO-STAGE LEAST SQUARES

JEL code: J16, C01

I. INTRODUCTION: A CONCEPTUAL FRAMEWORK

Over the past sixty years the attention devoted to the world of science has increased significantly, due to the recognition of its main role as one of the crucial determinants of the development of each country (Wagner, 2008). This is also one of the reasons why governments are continuously attempting to foster scientific research through the implementation of new incentive schemes (Shrum *et al*, 2007; Boyle, 2008; Frey 2003, 2009).

Following the pioneer works by Crane (1972) and Price (1986), social scientists began to investigate “The Republic of Science” (David, 2008:2) using appropriate tools of analysis which can help to explain its own patterns of knowledge creation and go beyond mere bibliometric indicators.

In particular, one may divide this huge research area into two-stream: the first (which we may label “bibliometric trend”) is focused on measuring the effectiveness and efficiency of research activity by taking into account inputs and, above all, outputs of the process (Moed, 2005; Weingart, 2005); the second (which we may label “relational trend”) analyses the causes and the scientific behaviours which may lead to different patterns of production, such as the scientists’ tendency to write in co-authorship with other scientists or the number of acknowledgments they include in their published articles (Hollis, 2001; Goyal, 2005; Goyal *et al*, 2006; Fachamps *et al*, 2006). This second stream of research has also introduced into this field many innovative tools of analysis, such as Social Network Analysis (Wasserman & Faust, 1994; Scott, 2000) and Textual Analysis (Roth and Cointet, 2010) alongside the most traditional ones.

This paper is assuming the relational trend as the starting point of analysis. The aim is to detect some of the factors which may affect researchers’ scientific productivity. In particular, the tendency to write papers in co-authorship will be used as a proxy of formal scientific collaboration amongst scientists, while the number of acknowledgements will be assumed as a proxy of the scientist’s ability to build informal collaboration networks. Both co-authorships and acknowledgments indices are interpreted as two of the main forces which could affect and drive scientific production, apart from individual talent.

One of the most common reasons which boosted economists to investigate the relationship between scientific productivity and collaborative behaviours (formal and informal) relies on the belief that collaborative behaviours might enhance both the quality and the quantity of publishable scientific articles. This represents also one of the most controversial assumptions researchers could assume, because of the antithetical results provided by applied works on the subjects (Hollis, 2001).

Owing to the fact that each scientific discipline has its own features which contradistinguish itself from any other, the scientific area of Geography has been selected even if (or just because of) the majority of the contributions already published focus on economics or hard sciences (physics and biology).

After having described the data collected and the criteria behind the construction of the dataset (§1), a two-stage least squares regression model is provided (§2) together with the interpretation of the results we obtained (§3).

II. DATA DESCRIPTION

Before proceeding with the estimation of the econometric model, it is necessary to specify both the rationale behind the dataset we built and the variables we may include in the model.

The original data come from *ISI Web of Science*¹, one of the largest bibliographic dataset provided by Thomson Reuters. The information we are interested in concerns the scientific discipline of Geography (which is also characterised by a strong interdisciplinarity with other social sciences, included Economics) and its pattern of publications during eight years (2000-2007). Therefore, all the bibliometric information regarding the first eight “Top Journals” have been downloaded. A scientific journal is classified as “Top” if its Impact Factor index (henceforth, IF) is ranked in the first 5 per each year. IF is calculated as follows:

$$IF = \frac{two-year_tot_citations_t_3}{tot_art_t_1 + t_2_A_journal} \quad (1)$$

Where $two-year_tot_citations_t_3$ is the amount of citations that the papers published in a generic “A journal” in a two-year period receive in the year following the publication; and $tot_art_t_1 + t_2_A_journal$ is the total number of articles published in such “A journal” in those two years.

Hence, IF rankings per each single year (2000-2007) were accessed in order to build the dataset. **Table 1** shows the annual IF ranking and the eight top journals which were selected.

¹ <http://www.thomsonreuters.com> [1 December 2010]

Table 1: Top Journals and Impact Factor

	2000	2001	2002	2003	2004	2005	2006	2007									
ID	JOURNAL	IF	rank	IF	rank	IF	rank	IF	rank								
TIBG	TRANSACTIONS OF THE INSTITUTE OF BRITISH GEOGRAPHERS	4.067	1	2.218	3	2.388	3	2.574	3	3.093	1	2.698	1				
GEC	GLOBAL ENVIRONMENTAL CHANGE - HUMAN AND POLICY DIMENSIONS	3.915	2	1.952	4	/	/	/	/	/	/	/	/				
PHG	PROGRESS IN HUMAN GEOGRAPHY	3.762	3	2.616	2	2.943	2	2.762	1	3.653	1	2.288	2	2.386	2		
AAAG	ANNALS OF THE ASSOCIATION OF AMERICAN GEOGRAPHERS	2.962	4	2.141	5	/	/	2.115	5	1.972	5	2.586	2	1.855	3	2.109	4
JEG	JOURNAL OF ECONOMIC GEOGRAPHY	2.679	5	2.519	4	3.222	1	3.139	1	/	/	/	/	/	/	/	/
EG	ECONOMIC GEOGRAPHY	/	/	/	/	1.757	5	2.325	4	/	/	2.455	4	/	/	1.909	5
PG	POLITICAL GEOGRAPHY	/	/	/	/	/	/	/	/	2.250	4	/	/	1.519	5	/	/
EPD	ENVIRONMENT AND PLANNING D-SOCIETY & SPACE	/	/	/	/	/	/	/	/	2.269	3	2.377	5	1.583	4	2.152	3

The sampling process originated a list of eight “top journals” which published a total of 2,474 articles (in the period 2000-2007) by 2,436 geographers (authors). To be reminded here that the dataset has been painstakingly cleaned in order to avoid misspelling problems and other trivial mistakes².

Additionally, we retrieved information about authors’ affiliation (university, department a/o research centre), city, country and number of acknowledgments per paper and per author. Moreover, in order to compute the productivity index, we looked for the total number of articles every single author has published from 1975 (which is the first year covered by ISI Web of Science) to 2007 (regardless IF).

The data concerning acknowledgments are also crucial in the analysis: they may be considered as a proxy of an unobservable variable concerning the authors’ ability to build informal collaboration networks. The total number of acknowledgments is 7,730, while the average number per paper is 3.12. **Table 2** summarises the information described above.

Table 2: Dataset Description

Years	8
Top Journals	8
Published articles	2,379
Total number of acknowledgments	7,730
Total number of acknowledged authors	636
Average number of acknowledgments per article	3.124
Average number of acknowledgments per author	3.110

Finally, we need to specify that 29 authors were dropped from the dataset due to the impossibility to assign a value to the variable gender³, while other 19 geographers were dropped due to nationality mismatches⁴. Therefore, the total number of observations became 2,379 instead of 2,436.

After having explained the dataset, a description of the variables is needed.

Amongst the data collected, eleven variables were taken into account in order to estimate their impact on the dependent variable, that is the logarithm of scientific productivity. Furthermore, it must be specified that, unless the data are collected per year, we are dealing with them as aggregate data. **Table 3** reports the variables description.

² Deriving, for example, from the inclusion or not of middle initials, “Mortimore, M.” appeared in ISI sometimes as “Mortimore M. L.” or as “Mortimore M. J. L.”. We opted for the latter option.

³ The information regarding gender was collected by online searching, because ISI Web of Science does not retrieve such type of data.

⁴ As gender, the information regarding the authors’ country was collected by online searching, because ISI Web of Science does not retrieve such type of data.

Table 3: Variables Description

VARIABLE	DESCRIPTION
<i>productivity</i>	index of productivity.
<i>coauthorship</i>	average number of co-authors per article.
<i>coauthinstab</i>	co-authors' instability index.
<i>moncoauth</i>	= 1 if the author writes <i>only</i> alone and = 0 otherwise.
<i>monpubl</i>	= 1 if the author writes only one article in the period 2000-2007 and = 0 otherwise.
<i>artop0007</i>	number of articles published in one or more of the eight "top journals" of the field (ISI Web of Science impact factor ranking) for the period 2000-2007.
<i>acknavg</i>	average number of acknowledgments per author.
<i>acknumb</i>	total number of acknowledgments per author.
<i>sciage</i>	author's scientific age.
<i>betweenness</i>	Social Network Analysis index of centrality in the authors' network.
<i>female</i>	= 1 if the author is female and = 0 otherwise.
<i>ukusal</i>	= 1 if the author's affiliation is either in England, Wales, North Ireland, Canada, USA, Australia or New Zealand; and = 0 otherwise.
<i>chijap</i>	= 1 if the author's affiliation is either in China or in Japan; and = 0 otherwise.

Firstly, the variable *productivity* is computed as the ratio between the total number of articles published in the eight top journals in the field of Geography for the period 2000-2007 and the author's scientific age. Instead, the author's scientific age is measured as the difference between the last year we are taking into account (2007) and the author's year of entry in ISI Web of Science database. These two adjustments allow us to control for the researchers' experience, without penalising young geographers.

Secondly, the variable *coauthorship* represents the average number of co-authors per article published by each scientist. More importantly, the variable *coauthinstab* is an index of the instability of the authors who a geographer writes with over time. This index is computed as the ratio between the total number of co-authors who always vary over the total number of co-authors per scientist. Furthermore, the variable *acknavg* represents the average number of acknowledgments per author, while *acknumb* is a measure of the total number of acknowledgments per scientist.

Thirdly, the variable *betweenness* is a Social Network Analysis (SNA)⁵ measure chosen amongst the available indices (**table 4**) of centrality in the authors' network⁶ (Freeman, 1979).

Table 4: SNA Indices of Network Centrality

	INDEX	AIM	COMMENT
Degree centrality	$C'_G = \frac{C_G}{(n-1)} = \frac{\sum_{i=1}^n l(a_i, a_k)}{(n-1)}$ <p>l = direct link between two adjacent nodes (a_i, a_k)</p> <p>C_G = un-normalised degree centrality</p>	It measures the number of links per each single node in the network.	The index is normalised according to the network magnitude.
Closeness centrality	$C'_P(a_k) = \frac{n-1}{\sum_{i=1}^n d(a_i, a_k)}$ <p>d = geodesic distance (the smallest one) between two nodes (a_i, a_k)</p>	It measures the node's strategic role in a network: it is related to its ability to reach each other node efficiently and autonomously.	The index is normalised according to the network magnitude.
Betweenness centrality	$C'_I(a_k) = \frac{2C_I(a_k)}{(n-2)(n-1)}$ <p>$\frac{(n-2)(n-1)}{2}$ = maximum value of $C_I(a_k)$</p>	It measures the potential strategic role which a node could play in the scientific collaboration network, through connecting sub networks which otherwise would be disconnected from each others.	<p>The index is normalised according to the network magnitude.</p> <p>The un-normalised index is computed as follows:</p> $b_{ij}(a_k) = \frac{g_{ij}(a_k)}{g_{ij}}$ <p>$b_{ij}(a_k)$ = proportion of geodesic distances (the smallest ones) between every two nodes in the network (i, j and $i < z < j$)</p> <p>$g_{ij}(a_k)$ = geodesic distance between i and j (where we also find z)</p> <p>g_{ij} = the shortest path between i and j</p>

Finally, the other independent variables are dummy variables which are computed as described in **table 3**.

⁵ For further information concerning this techniques of analysis, see Wasserman and Faust (1994) and Scott (2000).

⁶ This index has been computed using two different software for SNA: *Pajek 1.24* and *Ucinet 6*. Please refer to the bibliographical information for further details.

III. ECONOMETRIC ANALYSIS: TWO-STAGE LEAST SQUARES

The relationship we are interested in is the possible influence that co-authorship may have on productivity. Therefore, the crucial question to pose is whether or not writing with other authors may pay off. **Table 5** shows some descriptive statistics concerning the variables which we are going to include into the model.

Table 5: Descriptive Statistics

VARIABLE	MEAN	STANDARD DEVIATION	MINIMUM	MAXIMUM
<i>productivity</i>	.2307216	.2414975	0	2.1
<i>coauthorship</i>	2.563647	2.964842	0	28
<i>coauthinstab</i>	-.7805637	2.699442	0	1
<i>moncoauth</i>	2.143338	3.023049	0	28
<i>monpubl</i>	.7356032	.4411042	0	1
<i>acknavg</i>	3.365031	4.516346	0	34
<i>acknumb</i>	5.119798	8.164078	0	131
<i>sciage</i>	10.60614	8.66569	0	32
<i>betweenness</i>	.0012535	.0093117	0	.134
<i>female</i>	.2740647	.4461355	0	1
<i>ukusal</i>	.7545187	.4304627	0	1
<i>chijap</i>	.0281631	.1654734	0	1
Number of observations = 2,379				

The dependent variable is the logarithmic form of the productivity index, while we add all the remaining variables as independent, except for *artop0007*.

Additionally, the scientific age is included in its logarithmic form, in order to make the distribution closer to the normal:

$$\begin{aligned}
 \ln(\text{productivity}_i) = & \beta_0 + \beta_1 \text{coauthorship}_i + \beta_2 \text{coauthinstab}_i + \beta_3 \text{moncoauth}_i + \beta_4 \text{monpubl}_i + \\
 & + \beta_5 \text{acknavg}_i + \beta_6 \text{acknumb}_i + \beta_7 \ln(\text{sciage}_i) + \beta_8 \text{betweenness}_i + \beta_9 \text{female}_i + \\
 & + \beta_{10} \text{ukusal}_i + \beta_{11} \text{fem_ukusal}_i + \beta_{12} \text{chijap}_i + \beta_{13} \text{fem_chijap}_i + u_{it}
 \end{aligned} \tag{2}$$

As we can see from equation (2), two interaction terms have been added: one in order to control for the interaction between female and Anglo-Saxon countries, and the other in order to control for the interaction between female and China-Japan geographical area.

The main problem we are dealing with in this analysis is the presence of reverse causality: we may assume that cooperative patterns of publication could positively affect scientific productivity; but it could also be argued that productivity might affect co-authorship⁷.

Therefore, we are facing a problem of endogeneity of one of the explanatory variables (*coauthorship*), which could be correlated with the error term and therefore generate biased and inconsistent OLS estimators. The tool we are using to overcome this limit is the definition of an instrumental variable (henceforth, IV) and then estimating the model through a two-stage least squares (2SLS) regression.

Hence, we need to find a variable which satisfies two requirements: on one hand, it must not be correlated with u ; on the other hand, it must be correlated with *coauthorship*, which is in turn the instrumented dependent variable. One of the potential candidates is the variable *artop0007* (the total number of articles published on the top journals for the period 2000-2007). As suggested by Wooldridge (2009:512), in order to test our intuition it is allowed to regress the instrumented variable (*coauthorship*) on the IV (*artop0007*). The results of the regression gave evidence of a statistically positive correlation (1%) between the two variables.

Firstly, both a simple OLS⁸ and a 2SLS regressions have been run, in order to be able to test which of the two models could better explain the relationships we are testing.

In particular, we performed a Hausman test⁹: we tested against the null hypothesis that the OLS estimator was consistent and we obtained the result that the null hypothesis can be rejected at the 1% of significance. Hence, the two-stage method is preferred over the OLS regression. **Table 6** illustrates the results of both regressions.

⁷ For example, scientists could be more attracted by writing papers with authors who have an high productivity rather than with others with a low number of publications. It is well known that having a high number of publications improves authors' reputation inside academia.

⁸ Both the White test and the Breusch – Pagan test gave evidence for heteroscedasticity. We had to reject the null hypothesis of homoscedasticity with the 1% of confidence. Therefore, we corrected using robust standard errors. The full results are provided in the appendix.

White test: $\chi^2(81) = 687.68$, Prob > $\chi^2 = 0.000$

Breusch-Pagan test: $\chi^2(13) = 2,797.87$, Prob > $\chi^2 = 0.000$

⁹ We included the option “constant”, in order to include the constant term in the comparison of both the OLS and the two-stages estimates. Moreover, the option “sigma more” was added in order to use the same estimate of the variance of the error term for the two models. We are allowed to do so because the interpretation of the error term is exactly the same in both models we performed.

Table 6: OLS and Two-Stage Least Squares results (robust standard errors)

Dependent variable: log productivity

INDEPENDENT VARIABLES	OLS	2SLS
<i>coauthorship</i>	-.0375531*** (.0068257)	-1.072764*** (.2181598)
<i>coauthinstab</i>	.0080097*** (.0014219)	.0658599*** (.0127584)
<i>moncoauth</i>	.0353329*** (.0067771)	1.046485*** (.2136866)
<i>monpubl</i>	-.9503638*** (.0235148)	-2.604971*** (.3196528)
<i>acknavg</i>	-.0248672*** (.0029571)	.0062774 (.0054912)
<i>acknumb</i>	.0188264*** (.0027414)	-.0071947 (.0052597)
<i>Insciage</i>	-.8236862*** (.0047105)	-.8417227*** (.0169712)
<i>betweenness</i>	8.400692*** (1.1945)	27.35173*** (6.086655)
<i>female</i>	.0032183 (.0158871)	-.1204052 (.0932334)
<i>ukusal</i>	-.0127381 (.010743)	-.1619438** (.0775536)
<i>fem_ukusal</i>	.0000665 (.0191505)	.157393 (.1050781)
<i>chijap</i>	-.0188564 (.0272223)	.1106071 (.2892411)
<i>fem_chijap</i>	-.0643664 (.0475665)	.0934453 (.3709824)
<i>constant</i>	.4693892*** (.0265605)	2.371321*** (.3657218)
Observations	2,231	2,231
R-squared	0.9342	-
Instrumented variable	-	coauthorship
Instrument	-	artop0007

*** 1% level of significance

** 5% level of significance

Note: The quantities in parenthesis below the estimated coefficients refer to standard errors.

IV. RESULTS

As we can easily note, **table 6** does not report the R^2 for the 2SLS. This is an issue which is well documented in the literature (see amongst the others, Wooldridge, 2009:516-517)¹⁰.

Surprisingly, the variables related to the possible influence of the authors' ability to build informal networks of scientific collaborations (*acknavg* and *acknnumb*) are not significant at any of the usual confidence intervals. It can be assumed that having a rich group of informal collaborators does not affect the amount of published articles an author is able to write. Additionally, the variable *female* (which we add in order to test for the presence of gender discrimination) is not significant, even if we know from the data that only the 27.4% of geographers is composed by women. Consequently, also the coefficients of the interaction terms of female with the two geographical variables (*fem_ukusal* and *fem_chijap*) are not significant.

All the other variables are statistically significant at the 1% confidence level, except for *ukusal* which is significant at the 5%. From the regression emerges that the authors' propensity to write papers with other co-authors negatively affects their productivity. In particular, an increase of one unit of *coauthorship* produces a 10.7% decrease in authors' productivity. Moreover, the coefficient of coauthors' instability is in line with the result above. An increase of one point in the stability index raises productivity of the 6.5%. As a result, we might deduce that at least an author could write with co-authors, but he/she would better change them periodically, instead of keep on writing with them constantly. We can confirm our intuition by taking into account the variable *moncoauth*: authors who published *always* alone are the 10.4% more productive than the geographers who wrote at least one article in co-authorship.

Furthermore, being an "old" author (that is, an author with a high scientific age) negatively affects productivity (-.841). The geographical variables we added to test for the presence of different production patterns produced different results: on one hand, Anglo-Saxon countries present a significant coefficient; on the other hand China and Japan does not. We could also infer that *ukusal* may be a proxy of the presence of a possible language-bias in favour of the native English-speakers, but the issue needs more investigation.

Finally, the relational variable *betweenness* confirms the results related to the co-authorship coefficients. Since *betweenness* positively affects productivity, it should be the case that having few but strategic connections pays off; instead, a solid network of formal collaboration does not. It is also true that this SNA index is a measure of the authors' interposition amongst the different scientific networks; on one hand, only few but strategic collaborative connections could be enough in order to have a high index. On the other hand, being a "bridge" between

¹⁰ It is also possible to find useful comments on the issue in Stata website: <http://www.stata.com/support/faqs/stat/2sls.html> [10 January 2011].

authors with either different specialised knowledge or different research interests may be crucial (Piette & Ross, 1992).

V. CONCLUSION

The main purpose of this paper was to understand the relationship between scientific productivity and the scientists' tendency to collaborative behaviours. It also took into account some relational factors which could affect productivity related to formal and informal scientific networks. The data on the field of geography suggested that the informal influence (acknowledgments network) on productivity seems not to have any effect on productivity on one hand; co-authorship networks affect productivity in a negative way, but a variety of co-authors may increase geographers' productivity, on the other hand.

Finally, it emerged clearly a trade-off between the "transaction costs" from the collaboration and the need to mutual exchange of skills and knowledge complementarities which boost scientists to vary their co-authors.

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<http://www.thomsonreuters.com> [1 December 2010]

Appendices

a. OLS regression and tests for heteroskedasticity

1) OLS regression

```
regress lnproductivity coauthorship coauthinstab moncoauth monpubl acknavg
acknnumb lnsciage betweenness female ukusal fem_ukusal chijap fem_chijap
```

Source	SS	df	MS	Number of obs =	2231
Model	1317.22572	13	101.325056	F(13, 2217) =	2420.60
Residual	92.8025012	2217	.041859495	Prob > F =	0.0000
				R-squared =	0.9342
				Adj R-squared =	0.9338
				Root MSE =	.2046
Total	1410.02823	2230	.632299653		

lnproducti~y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
coauthorship	-.0375531	.0044279	-8.48	0.000	-.0462363 -.0288698
coauthinstab	.0080097	.0017611	4.55	0.000	.0045561 .0114634
moncoauth	.0353329	.0046667	7.57	0.000	.0261814 .0444844
monpubl	-.9503638	.0144456	-65.79	0.000	-.978692 -.9220355
acknavg	-.0248672	.0014906	-16.68	0.000	-.0277903 -.0219441
acknnumb	.0188264	.0008643	21.78	0.000	.0171313 .0205214
lnsciage	-.8236862	.0051767	-159.11	0.000	-.8338379 -.8135346
betweenness	8.400692	.4897773	17.15	0.000	7.440222 9.361162
female	.0032183	.0217239	0.15	0.882	-.039383 .0458196
ukusal	-.0127381	.0126187	-1.01	0.313	-.0374839 .0120077
fem_ukusal	.0000665	.0243759	0.00	0.998	-.0477355 .0478684
chijap	-.0188564	.0299824	-0.63	0.529	-.0776529 .0399402
fem_chijap	-.0643664	.0712033	-0.90	0.366	-.2039985 .0752657
_cons	.4693892	.0203067	23.11	0.000	.429567 .5092114

2) White test for heteroskedasticity

```
imtest, white
```

```
White's test for Ho: homoskedasticity
against Ha: unrestricted heteroskedasticity
```

```
chi2(81) = 687.68
Prob > chi2 = 0.0000
```

```
Cameron & Trivedi's decomposition of IM-test
```

Source	chi2	df	p
Heteroskedasticity	687.68	81	0.0000
Skewness	60.65	13	0.0000
Kurtosis	4.65	1	0.0310
Total	752.99	95	0.0000

3) Breusch-Pagan test for heteroskedasticity

estat hettest, rhs

Breusch-Pagan / Cook-Weisberg test for heteroskedasticity

Ho: Constant variance

Variables: coauthorship coauthinstab moncoauth monpubl acknavg acknumb
lnsciage betweenness female ukusal fem_ukusal chijap
fem_chijap

chi2(13) = 2797.87
Prob > chi2 = 0.0000

4) OLS regression (robust standard errors)

regress lnproductivity coauthorship coauthinstab moncoauth monpubl acknavg
acknumb lnsciage betweenness female ukusal fem_ukusal chijap fem_chijap,
vce(robust)

Linear regression

Number of obs = 2231
F(13, 2217) = 2969.83
Prob > F = 0.0000
R-squared = 0.9342
Root MSE = .2046

		Robust				
lnproductivity	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
coauthorship	-.0375531	.0068257	-5.50	0.000	-.0509386	-.0241676
coauthinstab	.0080097	.0014219	5.63	0.000	.0052213	.0107982
moncoauth	.0353329	.0067771	5.21	0.000	.0220427	.0486231
monpubl	-.9503638	.0235148	-40.42	0.000	-.9964772	-.9042503
acknavg	-.0248672	.0029571	-8.41	0.000	-.0306662	-.0190682
acknumb	.0188264	.0027414	6.87	0.000	.0134503	.0242024
lnsciage	-.8236862	.0047105	-174.86	0.000	-.8329238	-.8144487
betweenness	8.400692	1.1945	7.03	0.000	6.058237	10.74315
female	.0032183	.0158871	0.20	0.839	-.0279368	.0343734
ukusal	-.0127381	.010743	-1.19	0.236	-.0338055	.0083292
fem_ukusal	.0000665	.0191505	0.00	0.997	-.0374884	.0376213
chijap	-.0188564	.0272223	-0.69	0.489	-.0722403	.0345276
fem_chijap	-.0643664	.0475665	-1.35	0.176	-.157646	.0289132
_cons	.4693892	.0265605	17.67	0.000	.4173031	.5214752

b. Two-stage least squares regression and Hausman test

1) 2SLS regression

```
ivregress 2sls lnproductivity coauthinstab moncoauth monpubl acknavg
acknnumb lnsciage betweenness female ukusal fem_ukusal chijap fem_chijap
(coauthorship=artop0007)
```

```
Instrumental variables (2SLS) regression                Number of obs =    2231
                                                       Wald chi2(13) = 1295.00
                                                       Prob > chi2    = 0.0000
                                                       R-squared     = .
                                                       Root MSE     = 1.033
```

lnproductivity	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
coauthorship	-1.072764	.1346329	-7.97	0.000	-1.336639 - .808888
coauthinstab	.0658599	.0115808	5.69	0.000	.0431619 .0885579
moncoauth	1.046485	.1318015	7.94	0.000	.7881584 1.304811
monpubl	-2.604971	.2243854	-11.61	0.000	-3.044759 -2.165184
acknavg	.0062774	.0085204	0.74	0.461	-.0104223 .0229772
acknnumb	-.0071947	.0054939	-1.31	0.190	-.0179625 .0035731
lnsciage	-.8417227	.0262399	-32.08	0.000	-.8931521 -.7902934
betweenness	27.35173	3.467347	7.89	0.000	20.55586 34.14761
female	-.1204052	.1108268	-1.09	0.277	-.3376217 .0968113
ukusal	-.1619438	.0665252	-2.43	0.015	-.2923308 -.0315568
fem_ukusal	.157393	.12472	1.26	0.207	-.0870538 .4018398
chijap	.1106071	.1522932	0.73	0.468	-.1878821 .4090962
fem_chijap	.0934453	.3600844	0.26	0.795	-.6123072 .7991977
_cons	2.371321	.2645924	8.96	0.000	1.85273 2.889913

```
Instrumented: coauthorship
Instruments: coauthinstab moncoauth monpubl acknavg acknnumb lnsciage
betweenness female ukusal fem_ukusal chijap fem_chijap artop0007
```

2) Hausman test

```
predict ivresid, residuals (148 missing values generated)
```

```
est store ivreg
```

```
hausman ivreg ., constant sigmamore df(1)
```

Note: the rank of the differenced variance matrix (1) does not equal the number of coefficients being tested (14); be sure this is what you expect, or there may be problems computing the test. Examine the output of your estimators for anything unexpected and possibly consider scaling your variables so that the coefficients are on a similar scale.

	---- Coefficients ----			
	(b)	(B)	(b-B)	sqrt(diag(V_b-V_B))
	ivreg	.	Difference	S.E.
coauthorship	-1.072764	-.0375531	-1.035211	.0262943
coauthinstab	.0658599	.0080097	.0578502	.0014694
moncoauth	1.046485	.0353329	1.011152	.0256832
monpubl	-2.604971	-.9503638	-1.654608	.042027
acknavg	.0062774	-.0248672	.0311446	.0007911
acknnumb	-.0071947	.0188264	-.0260211	.0006609
lnsciage	-.8417227	-.8236862	-.0180365	.0004581
betweenness	27.35173	8.400692	18.95104	.4813561
female	-.1204052	.0032183	-.1236235	.00314
ukusal	-.1619438	-.0127381	-.1492057	.0037898
fem_ukusal	.157393	.0000665	.1573265	.0039961
chijap	.1106071	-.0188564	.1294634	.0032884
fem_chijap	.0934453	-.0643664	.1578117	.0040084
_cons	2.371321	.4693892	1.901932	.048309

 b = consistent under Ho and Ha; obtained from ivregress
 B = inconsistent under Ha, efficient under Ho; obtained from regress

Test: Ho: difference in coefficients not systematic

chi2(1) = (b-B)'[(V_b-V_B)^(-1)](b-B)
 = 1550.01
 Prob>chi2 = 0.0000
 (V_b-V_B is not positive definite)

3) 2SLS regression (robust standard errors)

ivregress 2sls lnproductivity coauthinstab moncoauth monpubl acknavg
 acknumb lnsciage betweenness female ukusal fem_ukusal chijap fem_chijap
 (coauthorship=artop0007), vce(robust)

Instrumental variables (2SLS) regression
 Number of obs = 2231
 Wald chi2(13) = 3173.21
 Prob > chi2 = 0.0000
 R-squared = .
 Root MSE = 1.033

lnproducti~y	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
coauthorship	-1.072764	.2181598	-4.92	0.000	-1.500349	-.6451783
coauthinstab	.0658599	.0127584	5.16	0.000	.0408539	.0908659
moncoauth	1.046485	.2136866	4.90	0.000	.6276664	1.465303
monpubl	-2.604971	.3196528	-8.15	0.000	-3.23148	-1.978463
acknavg	.0062774	.0054912	1.14	0.253	-.0044852	.01704
acknumb	-.0071947	.0052597	-1.37	0.171	-.0175036	.0031142
lnsciage	-.8417227	.0169712	-49.60	0.000	-.8749857	-.8084598
betweenness	27.35173	6.086655	4.49	0.000	15.42211	39.28136
female	-.1204052	.0932334	-1.29	0.197	-.3031393	.0623289
ukusal	-.1619438	.0775536	-2.09	0.037	-.313946	-.0099416
fem_ukusal	.157393	.1050781	1.50	0.134	-.0485563	.3633423
chijap	.1106071	.2892411	0.38	0.702	-.4562951	.6775092
fem_chijap	.0934453	.3709824	0.25	0.801	-.6336669	.8205575
_cons	2.371321	.3657218	6.48	0.000	1.65452	3.088123

Instrumented: coauthorship

Instruments: coauthinstab moncoauth monpubl acknavg acknumb lnsciage betweenness
 female ukusal fem_ukusal chijap fem_chijap artop0007