WORKER REMITTANCES IN GROWTH REGRESSIONS: THE PROBLEM OF COLLINEARITY

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***Abstract*.** The sign of worker remittances in growth regressions is heavily disputed in the literature. Comparing two growth regressions with different signs for the remittance variable we show that collinearity with the lagged dependent variable might indicate that collinearity should be investigated comprehensively and might lead to a change in specifications, which differ in the variance inflation factors (VIF). In our case the variance inflation factor for remittances depends on the use of a five or one-year lag of the lagged dependent. In the regression with a VIF below ten, the standard critical value, the sign of remittances is positive.

1. Introduction

Every researcher who has suggested one of the about 140 regressors in growth regressions (Durlauf et al. 2005) had to respond to the question of reversed causality. It is much less clear though that anyone had to respond to the question whether or not the sign of a suggested regressor may be turned around by approximate collinearity (see Davidson McKinnon 2004 on the basics) with the lagged dependent variable although there were some debates on the right signs in growth regression. In general, the issue is important because each chapter in any textbook on development economics suggests the relevance for income effects. If income effects are important collinearity with the lagged dependent variable can be a major issue in any growth regression. Examples are controversies in regard to applications for development aid (see Doucouliagos, H. and M**.** Paldam (2008) and on the impact of worker remittances on growth found by Chami et al. (2005). It is this latter case in which we are interested in relation to the collinearity issue.

   Chami et al. (2005) have argued that remittances provide an incentive to reduce effort thereby making weak economic performances more likely. They find negative impacts of remittances on growth in a cross-section regression. In Lucas (2005) and IMF (2005) this result is attributed intuitively to weak or inadequate instruments and in the latter no growth effect is found. Catrinescu et al. (2009) extend the approach of Chami et al. to include policy and institutional variables and estimate a panel using the Anderson-Hsiao estimator. They find some significantly positive results for the impact of remittances on growth, but these are reported to be not very robust. Giuliano and Ruiz-Arranz (2008) add remittances multiplied to financial variables as a regressor and find positive growth effects for financially less developed countries. In summary, these papers see the reason for the negative sign found by Chami et al. (2005) in inadequate instruments, omitted variables and inadequate estimation methods. We advance another possibility that might

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be useful for future research: approximate collinearity with the lagged dependent variable.

2. Methodology
Growth regressions can be written as follows (Durlauf et al. 2005).

$$Log(y_t) = \alpha_i + (\beta+1)log(y_{t-1}) + \gamma x_1 + \eta x_2 + u_{it}$$

'log' indicates a natural logarithm, y is GDP per capita or per worker, x_1 denotes regressors used in mathematically formulated growth models, in particular the augmented Solow model (see Mankiw et al. 1992) and x_2 denotes other regressors, which are added although they are not included in a growth model. Examples for the latter are official development aid or worker remittances. Such effects are normally interpreted to mirror the impact of a variable on the total factor productivity (see Rogriguez (2006)), which can be considered to be a weighted average of sectoral productivities. These variables then either affect the weights of the sectors through the shift of demand and factor inputs or they have an impact on the sectoral technical progress (see Timmer and Szirmai (2000)). The expected signs for the coefficient of the lagged dependent variable normally obtained in growth regressions are $\beta < 0 < \beta+1$.

   We will estimate such a growth regression for more than 40 countries with per capita income above \$1200 in prices of the year 2000.[1] In the first instance we obtain the result that the impact of worker remittances on growth is negative under some additional assumptions. One of these assumptions is the use of a lagged dependent variable with a five years lag that is significantly correlated with the remittance variable. However, a one-year lagged dependent variable is much less significantly correlated with remittances, depending on the set of controls used though. Using a one-year lagged dependent variable and reworking the regression towards having only significant variables the sign for remittances changes into a positive one. Finally, the variance inflation factors (VIF) of all regressors[2] in both equations are calculated indicating that the remittance variable in the second equation is much less correlated with other regressors than in the first equation. Due to other multi-collinearities it remains an open question, which of the changes is actually turning the sign around, but it is the one with the lagged dependent variable which is economically plausible, therefore checked first and easily tested before the ultimate plausibility comes from the comparison of variance inflation factors indicating the strength of the multicollinearity.

   All data are taken from the World Development Indicators. We use the fixed effects method, which is known to have a downward bias for the lagged dependent variable of an order of magnitude of *1/T,* if we have more than thirty observations as we do in the first regression. If we have less than thirty observations we use the systems GMM method of Arellano-Bover (1995) because fixed effects estimation is then underestimating the coefficient of the lagged dependent variable. With this method we can use instruments to correct for the endogeneity of the lagged dependent variable and other regressors. In our

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[1] In related work on countries with per capita income below \$1200 we found no ambiguity in the coefficients for aid and remittances.
[2] See Kennedy (2003) for an extensive treatment.

case the coefficient of the lagged dependent variable is slightly larger than that of the fixed effects regression and the test for the validity of the instruments and not having too many of them (see Roodman 2007) is also passed.

3. Results

The growth regression for the log of the GDP per capita, *log(gdppc)*, we would have defended in the first instance is as follows (p-values in parentheses)[3].

Log(gdppc) - log(gdppc(-5)) = -4.66-0.14log(gdppc(-5) + 0.11log(gfcfgdp)
                    (0.0003)   (0)   (0)

-.0245log(gfcfgdp(-5)) + 0.0014Lit(-1) +1.86(1/t) + 6.54(wr/gdp)^2 -1.57wr/gdp
(0.05)   (0.092)   (0.0125)   (0.0003)   (0.0004)

+ 1.08wr(-1)/gdp(-1) – 3.76(wr(-1)/gdp(-1))^2+ 0.52oda/gdp -2.78(oda/gdp)^2 +
 (0.015)   (0.004)   (0.03)   (0.027)

+ 0.2log(wld) -0.057 log(l)   (1)
 (0)   (0.046)

Periods: 34 (1971 2005). Countries: 45. Obs.: 634. Adj.R^2 =0.996; DW=1.77

The lagged dependent variable has a sign and size of the coefficient in accordance with the expectation given above. The sum of the coefficients of the investment variables, *gfcfgdp*, is positive. Literacy, *Lit*, also has a positive sign and the growth of the GDP of the world, *wld*, as an income argument in the export demand function stemming from the idea of growth modeled with imported inputs in Bardhan and Lewis (1970) has a positive sign. The natural logarithm of the labour force, *log(l)*, has a negative sign of approximately the same order of magnitude as the world income variable.[4] The squared values for remittances, *wr/GDP*, and aid, *oda/GDP*, are very small. Therefore the linear ones dominate. Under the assumption that variables and their lags are of similar size remittances have a negative impact and aid has a positive one. However, a look at Table 1 shows that the regressors with the exception of the lagged investment variables are pairwise significantly correlated with the lagged dependent variable. The result may therefore stem from collinearity, which may have an impact on the sign of regressors. Table 2 shows results from regressing the remittance variables on the GDP per capita and its one and five year lags. The correlation is most strong for the five year lag used in the

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[3] A value of (0) indicates zeros for four digits. Three lagged growth rates are employed as serial correlation correction.

[4] Using the formulas in Mutz and Ziesemer (2008) and assuming an elasticity of production for capital of 0.33, we get a price elasticity of export demand of (-6.9) in regression (1) and (-4) in regression (2) below. Both values seem quite reasonable. Again according to these growth rate formulas we can obtain the income elasticity of export demand as the ratio of the coefficients for the world income and the labour variable. This coefficient is 3.5 for the first regression and therefore far too high and slightly above unity for the second regression.

above regression. One may therefore want to avoid five year lags. Changing them into one-year lags and eliminating the most highly insignificant regressors lead us to the following result.

$$d(\log(gdppc)) = c - 0.09\log(gdppc(-1)) + 0.123\log(gfcfgdp) - 0.09\log(gfcfgdp(-1)) \quad (2)$$
$$\quad\quad (0) \quad\quad\quad\quad\quad (0) \quad\quad\quad\quad\quad (0)$$

$$- 0.19d(\log(L)) + 0.00146sum(Lit) + 1.06(wr(-1)/gdp(-1))^2 + 0.3oda(-1)/gdp(-1)$$
$$\quad (0.015) \quad\quad\quad (t=1.67)) \quad\quad\quad (0.073) \quad\quad\quad\quad\quad (0.033)$$

$$- 0.52(oda(-1)/gdp(-1))^2 + 0.114\log(wld) - 0.099\log(l)$$
$$\quad (0.052) \quad\quad\quad\quad\quad (0.0001) \quad\quad (0.0003)$$

Per.: 23 (1981 2005). Countr.: 42. Obs.: 558. s.e.e.: 0.037. J=267. Instr.rank:257. p(J) = 0.168.

For the literacy variable we now use a polynomial distributed lag of the first degree with 10 lags, which has negative growth effects for the first five lags and but significantly positive effects thereafter. These lags cost us some observations and therefore the adequate method is that of Arellano-Bover (1995). Moreover, the aid variables are used now with a one-year lag. The major difference though is that the remittance variable now has a positive effect, which it did not when using the five-year lag for the lagged dependent variable. Moreover, only the squared lag of remittances is significant.

   The econometric literature on multicollinearity emphasizes the variance inflation factor, $1/(1-R_i)$, where R_i is the coefficient of determination for the regression of regressor i on all the other regressors. In Table 3 we provide the values for R_i and the variance inflation factors for both regressions.[5] The worker remittance variables have a high collinearity in the first regression but a much lower one in the second.

4. Conclusion
In both regressions the aid variable has a significantly positive sign.[6] The other variables have the expected sign. Switching from the five-year lag to the one-year lag because of the collinearity with the lagged dependent variable in the first regression ultimately changes the sign of the remittance variable after other adjustments are made. The variance inflation factors indicate that the sign and significance of the remittance variable are based on correlation with the other regressors in the first equation but much less so in the second, where the variance inflation factor is below the standard critical value of 10 (see Kennedy 2003). Therefore we cautiously suggest that the positive sign for remittances is more convincing for our sample. As a tentative interpretation, remittances and aid are unlikely to contribute to total factor productivity growth (tfp) via technical change; but rather remittances and aid seemingly are spent in sectors with above average tfp and thereby shift more weight to them and generate higher aggregate tfp levels.

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[5] For this purpose we use the fixed effects version of equation (2) because we do not have lagged dependent variables in most cases and we need an R-squared value and therefore a constant; both are not calculated in the Arellano-Bover method.

[6] The squared term generates a function with peaks at 9.35% and 28.8% of GDP only.

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Appendix 1: Tables

Table 1: Uncontrolled correlation matrix and marginal significance levels

Covariance Analysis: Ordinary
Sample (adjusted): 1971 2005
Included observations: 650 after adjustments
Balanced sample (listwise missing value deletion)
Correlation

Probability	1	2	3	4	5	6	7	8	9	10	11	12	13
1.LOG(GDPPC)	1												

2.LOG(GDPPC(-1))	1.00	1.00											
	0.00	-----											
3.LOG(GDPPC(-5))	0.98	0.98	1.00										
	0.00	0.00	-----										
4.LOG(GFCFGDP)	0.00	-0.02	-0.10	1.00									
	0.99	0.66	0.01	-----									
5.LOG(GFCFGDP(-5))	0.04	0.04	0.03	0.44	1.00								
	0.27	0.32	0.40	0.00	-----								
6.LIT(-1)	0.56	0.56	0.54	-0.12	-0.12	1.00							
	0.00	0.00	0.00	0.00	0.00	-----							
7.D(LOG(L))	-0.09	-0.08	-0.07	-0.05	0.01	-0.15	1.00						
	0.02	0.04	0.09	0.22	0.75	0.00	-----						
8.(WR/GDP)^2	-0.19	-0.19	-0.19	0.18	0.21	-0.10	0.09	1.00					
	0.00	0.00	0.00	0.00	0.00	0.01	0.02	-----					
9.WR/GDP	-0.27	-0.27	-0.26	0.16	0.17	-0.23	0.08	0.94	1.00				
	0.00	0.00	0.00	0.00	0.00	0.00	0.04	0.00	-----				
10.WR(-1)/GDP(-1)	-0.26	-0.26	-0.26	0.15	0.19	-0.23	0.08	0.92	0.98	1.00			
	0.00	0.00	0.00	0.00	0.00	0.00	0.05	0.00	0.00	-----			
11.(WR(-1)/GDP(-1))^2	-0.19	-0.19	-0.19	0.15	0.22	-0.09	0.08	0.95	0.91	0.94	1.00		
	0.00	0.00	0.00	0.00	0.00	0.02	0.04	0.00	0.00	0.00	-----		
12.ODA/GDP	-0.41	-0.41	-0.39	0.06	0.09	-0.30	0.09	0.62	0.62	0.62	0.62	1.00	
	0.00	0.00	0.00	0.11	0.02	0.00	0.02	0.00	0.00	0.00	0.00	-----	
13. (ODA/GDP)^2	-0.28	-0.28	-0.26	0.12	0.13	-0.17	0.08	0.56	0.52	0.52	0.55	0.91	1.00
	0.00	0.00	0.00	0.00	0.00	0.00	0.04	0.00	0.00	0.00	0.00	0.00	-----

Table 2: Collinearity of remittance and GDP per capita

Regressors	Coefficient	Std. Error	t-Statistic	Prob.
Dependent Variable: WR/GDP				
C	0.034	0.085	0.393	0.694
LOG(GDPPC)	0.038	0.027	1.437	0.151
LOG(GDPPC(-1))	-0.018	0.024	-0.743	0.458
LOG(GDPPC(-5))	-0.020	0.012	-1.690	0.091
Dependent Variable: (WR/GDP)^2				
C	0.000	0.019	0.003	0.998
LOG(GDPPC)	0.005	0.010	0.467	0.641
LOG(GDPPC(-1))	0.001	0.009	0.123	0.902
LOG(GDPPC(-5))	-0.005	0.004	-1.388	0.166
Dependent Variable: WR(-1)/GDP(-1)				
C	0.037	0.088	0.415	0.679
LOG(GDPPC)	0.038	0.028	1.363	0.173
LOG(GDPPC(-1))	-0.012	0.025	-0.468	0.640
LOG(GDPPC(-5))	-0.026	0.012	-2.166	0.031
Dependent Variable: (WR(-1)/GDP(-1))^2				
C	0.004	0.021	0.181	0.857
LOG(GDPPC)	-0.002	0.010	-0.181	0.857
LOG(GDPPC(-1))	0.010	0.009	1.069	0.285
LOG(GDPPC(-5))	-0.008	0.004	-1.863	0.063

Table 3: Variance Inflation Factors

	R-sq. Regr.1	R-sq. Regr.2	*VIF Regr. 1*	*VIF Regr.2*
LOG(GDPPC(-1))	-	0.984	-	63.0
LOG(GDPPC(-5))	0.978	-	46.4	-
LOG(GFCFGDP)	0.657	0.784	2.9	4.6
LOG(GFCFGDP(-1))	-	0.805	-	5.1
LOG(GFCFGDP(-5))	0.656	-	2.9	-
LIT(-1)	0.981	-	52.7	-
1/(@trend)	0.922	-	12.9	-
LOG(GDPPC(-1))-LOG(GDPPC(-6))	0.898	-	9.8	-
LOG(GDPPC(-2))-LOG(GDPPC(-7))	0.938	-	16.0	-
LOG(GDPPC(-3))-LOG(GDPPC(-8))	0.881	-	8.4	-
(WR/GDP)^2	0.985	-	65.2	-
WR/GDP	0.992	-	128.8	-
WR(-1)/GDP(-1)	0.992	-	130.0	-
(WR(-1)/GDP(-1))^2	**0.985**	**0.849**	**66.4**	**6.6**
ODA/GDP	0.960	0.962	24.9	26.5
(ODA/GDP)^2	0.942	-	17.3	-
ODA(-1)/GDP(-1)	-	0.962	-	26.5
(ODA(-1)/GDP(-1))^2	-	0.914	-	11.7
LOG(WLD)	0.964	0.906	28.1	10.6
LOG(L)	0.998	0.999	604.6	869.6
d(log(L))	-	0.389	-	1.6