CO₂ intensity and GDP per capita

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Abstract

The relationship between CO_2 intensity and GDP per capita is studied. Most rich countries show falling CO_2 intensity over time and a negative correlation with GDP per capita. Many poor and medium rich countries show the opposite, a positive time trend and a positive correlation with GDP per capita. For about a half of the countries with a negative correlation between CO_2 intensity and GDP per capita, and in particular the largest economies of the world, there is strong evidence that CO_2 intensity falls at a diminishing rate as countries get richer. Hence, economic growth will not by itself go very far in reconciling economic growth and reductions in CO_2 emissions. There are indications that poor and medium rich countries experience a boost in CO_2 intensity as they embark on industrialization. This will also make it harder to reconcile economic growth and cuts in CO_2 emissions. Falling CO_2 intensity of GDP in rich countries may not give a correct indication of what happens to world GDP because rich countries have outsourced much of their energy-intensive production to poorer countries. Looking at the production of eleven minerals 1960-2017, we find little or no evidence of dematerialization. Production of five minerals has grown more rapidly than world GDP and that of four other minerals more rapidly than world population.

Keywords: carbon dioxide; economic growth; CO₂ intensity; mineral production; GDP; economic growth; dematerialization

JEL classification: N50, O13, Q32, O44, Q43, Q54

1. Introduction

Is economic growth compatible with reduction in carbon dioxide emissions? If so, carbon dioxide emissions per unit of GDP (hereafter CO_2 intensity) will have to fall. New technologies for energy production on a grand scale are likely to be necessary for this, but it would also help if there are structural trends accompanying economic growth that would bring the CO_2 intensity down. This is not unlikely, as economic growth is accompanied by disproportionate growth of services, which are less energy intensive than material production (Medlock and Soligo, 2001).

What is the historical record? As a part of its battery of world economic indicators, the World Bank publishes carbon dioxide content per unit of GDP at fixed prices for most countries in the world. In this paper we use this data to investigate the historical record across countries and, in particular, how CO_2 intensity is related to GDP per capita. We get mixed results, and yet a tendency that the CO_2 intensity falls as countries get richer, but at a diminishing rate.

According to the BP Statistical Review of World Energy, more than 80 percent of primary commercial energy still comes from fossil fuels. Since most CO₂ emissions are caused by burning fossil fuels, what has happened to CO₂ intensity is in large measure a reflection of what has happened to energy intensity. Many papers on that subject have been published, and most indicate that energy intensity falls as GDP per capita increases, or that the relationship has an inverted U-shape. Csereklyei, Rubio-Varas and Stern (2016) find, for a sample of 99 countries, that energy intensity falls as countries grow richer, but point out that energy intensity may increase in countries experiencing no growth. They also point out that the increasing energy intensity often observed for poor countries could be due to a transition from non-commercial biomass energy to commercial energy. They include non-commercial energy in their data, but recognize the unreliability of such data. Most other studies use only commercial energy. Medlock and Soligo (2001) find the inverted U-shape for intensity of commercial energy, for a panel of 28 countries.

In a recent paper, Semieniuk (2018) investigates the "green growth hypothesis", that is, whether a faster development in productivity will reduce the energy intensity of the economy. Using a large but unbalanced panel of 180 countries 1950-2014 he finds that faster growth is not greener; a higher rate of labor productivity growth is typically associated with a higher rate of growth of energy input per unit of labor, canceling the effect on energy intensity. Hence, faster productivity growth will not contribute to reconciling economic growth and reduction in CO_2 emissions.

Two papers study the relationship between CO_2 emissions and GDP. Bella, Massidda and Mattana (2014) study the relationship between total CO_2 emissions and total GDP for a panel of 22 OECD-countries. They find an inverted U-shape for most countries, which most likely implies a similar shape as well for CO_2 intensity and GDP per capita, as for most countries GDP and GDP per capita have moved in the same direction. Jakob et al. (2012) study the growth of CO_2 emissions and GDP for a sample of 51 countries. They break their sample into developing and industrialized countries and find that both grew at a rate higher than average in developing countries while there is no significant relationship between the growth rates of GDP and the use of energy for developed countries. These results are not directly comparable to ours, but neither do they contradict them.

Looking at the relationship between CO_2 intensity, or energy intensity, and GDP per capita implies that a structural change in GDP as countries grow richer is seen as a driver of changes in CO_2 emissions or energy use. A rationale has already been advanced; as countries get richer, more and more of presumably less energy intensive services is produced and CO_2 intensity falls, while in countries just beginning their industrialization the opposite might happen. But things are more complicated than that. Energy or CO_2 intensity might fall with no change in GDP per capita because of technological progress leading to increased energy efficiency across economic sectors or a transition from fossil fuels to other energy sources, or even between different fossil fuels (such as less reliance on coal and greater use of natural gas). Two studies of the US economy try to tease out how much of energy savings is due to increased energy efficiency (better technology) and how much is due to structural changes following changes in GDP per capita. Metcalf (2008) found that most of the reduction in energy intensity is due to improvements in energy efficiency while Huntington (2010) came to the opposite result. As pointed out by Huntington, the difference could be due to the degree of disaggregation in the data. So, to analyze this question, one needs not only country-specific disaggregated data, but the level of aggregation could have a critical bearing on the answer.

There are more devils in the details. In a recent paper, Croner and Francovic (2018) study structural versus efficiency factors behind changes in energy intensity, using detailed input-output coefficients for a number of countries. They point out that production-based data give more importance to structural factors than consumption-based data would do, because rich countries have to a large extent outsourced the production of CO₂-intensive goods to developing countries, a point also made by Dieter Helm (2012) with the British economy as an example. In the last part of the paper we investigate the development in the extraction of eleven minerals important for industrial production. This is important because digging up minerals and converting them into useful products is energy-intensive. Since the world economy is still critically dependent on fossil fuels, this has implications for reducing carbon dioxide emissions. Unless extraction and processing of minerals can quickly be based on using energy from renewable resources, it will continue to contribute to emissions of greenhouse gases, especially carbon dioxide.

2. How CO₂ emissions have evolved

Figure 1 shows the development of the CO_2 intensity world wide, for real GDP measured in 2010 US dollars. The CO_2 intensity fell steadily from 1960 to 2000 and stagnated after that. This is curious, as efforts to develop green energy and otherwise reduce carbon dioxide emissions have been particularly strong after 2000. When China is removed from the sample the stagnation disappears. Nevertheless, the CO_2 intensity has fallen more slowly for the world excluding China in this century than it did before, so we still face the paradox why efforts at decarbonization have achieved so little since they appeared on the world agenda.



Figure 1: World CO_2 emissions (kg per 2010 US\$ of GDP) 1960-2014 with and without China. Data from the World Bank.

In the World Bank sample there are 190 countries, but not all are represented for the entire 1960-2014 period (the countries and the years they are represented are listed in the Appendix). Looking at the time trend of CO_2 intensity across countries, we find that most of 36 countries with a GDP per capita of more than 23,000 dollars have a negative time trend, but for five the trend coefficient is insignificant. Below 23,000 dollars of GDP per capita a significantly positive time trend begins to show up, and then we are down to what may be termed medium rich countries; the richest ones of those with a positive time trend are Greece, Portugal and Saudi Arabia. For the remaining 154 countries, which may be characterized as medium rich or poor, we get a significantly positive time trend for about a half (69), while for 58 we get a significantly negative time trend, and for 28 we get no significant trend at all. The CO_2 intensity has thus tended to rise rather than fall for medium rich and poor countries, contrary to what has happened in rich countries.

CO₂ intensity and GDP per capita

One reason why the CO_2 intensity has been falling over time in many countries is that GDP per capita has been increasing. If CO_2 intensity falls as GDP per capita increases, for reasons already mentioned, this will show up as a falling time trend of CO_2 intensity. We now turn to investigating the relationship between CO_2 intensity and GDP per capita. We focus attention on countries with a negative relationship between these two and ask in particular whether there is a tendency for CO_2 intensity to fall at a declining rate when countries get richer, as Figure 1 indicates might be the case.

The results of a regression analysis of individual countries are summarized in the Appendix. We get a significantly negative correlation between CO_2 intensity and GDP per capita for 94 countries, a significantly positive one for 48, while 48 coefficients are insignificant. The negative correlation is most prominent for rich countries. Ordering countries by GDP per capita, the first countries to appear with a positive correlation are Greece and Portugal, and thereafter positive correlation begins to appear with an increasing frequency. The countries with a negative time trend and a negative correlation with GDP per capita are mostly the same. There are far fewer countries with a positive correlation between CO_2 intensity and GDP per capita (48) than those with a significantly positive time trend (69), so there are more countries with no significant correlation between CO_2 intensity and GDP per capita (48) than those with a significantly positive time trend (48) than those with a negative time trend (48) than those with a negative time trend (48) than those with an insignificant time trend (33).

Does CO_2 intensity fall with GDP per capita at a diminishing rate? To investigate this, we regressed CO_2 intensity on GDP per capita and GDP per capita squared, for the 94 countries that showed a negative relationship between CO_2 intensity and GDP per capita. The results are shown in the Appendix. If CO_2 intensity falls at a diminishing rate, the coefficient of the GDP-squared term should be positive. We get a significantly positive coefficient for about a third of the countries (35), and a positive but not significant coefficient for another third (28). For ten countries both regression coefficients are negative but insignificant, giving no support to the diminishing decline of the CO_2 intensity as countries become richer.

Then there are 21 cases where the regression coefficient of GDP per capita is positive while that of the squared GDP-term is negative. In that case, the CO_2 intensity would increase with GDP per capita up to a certain level and fall thereafter. A pattern like that is consistent with energy intensity of economies increasing as countries industrialize and become richer, but as they pass a certain point it falls. Does it do so at a diminishing rate? To investigate that we add a cubic term of GDP per capita to the regression for those countries. The results are shown in the Appendix. For a little over a half (13) of those countries we get a significantly positive regression coefficient, indicating that in the decline phase the CO_2 intensity does indeed fall with a rising GDP per capita at a diminishing rate and might even rise. For five countries we get positive but insignificant coefficients, but for three we get negative but insignificant coefficients.

Hence there appears to be strong support for the hypothesis that CO_2 intensity falls with GDP per capita at a diminishing rate. We find significant statistical support for this for about a quarter of the world's countries and a weak (right sign but insignificant coefficient) support for about 30 others. More importantly, the strongest support is provided by the richest and largest economies in the world; the United States, China, Germany, the United Kingdom, and many others. But even some poor countries also follow this pattern.

A potential explanation of why the CO_2 intensity of world GDP has fallen more slowly after 2000 is that many enough countries may have reached the level of GDP per capita where further gains in declining CO_2 intensity are small. While many countries are still so poor that they are unlikely to have reached that level, what happens in rich countries, which are responsible for most CO_2 emissions, may be decisive.

An illustration

It would require too much space to illustrate the modeling results for all countries, but it is of interest to compare them with the actual development in the largest economies of the world. After all, these countries have, by their sheer size, most effect on world GDP and also on world emissions of CO_2 , even if the CO_2 intensity of GDP varies considerably between them (the CO_2 intensity of China's GDP is about four times that of the United States). Figure 2 shows the development of the CO_2 intensity for the eight countries with the highest total GDP in 2014 and compares it with our modeling results. The model reproduces the actual development in the United States, Germany, France and the United Kingdom quite well (note that we only have data from Germany after 1991). The result is less good for China; in that country the CO_2 intensity has had a rickety ride, with a rapid fall in the 1960s, then a rise, and a fall again from the late 1970s. For China we use a function including both squared and cubed GDP per capita (see Appendix). The modeled CO_2 intensity stays fairly flat for the first two decades. Then, as GDP begins to increase, the negative squared term produces a fall in the CO_2 intensity, but in the last years the positive cubic term weighs more heavily and actually produces an increase instead of merely slower decline or stagnation, so obviously this term overcompensates.¹

The model for Japan, which also has a cubic term, fits rather well. The CO_2 intensity of the Japanese GDP increased to the mid-1970s and has fallen thereafter. The model captures this quite well. The negative squared term produces a decline in CO_2 intensity after the mid-1970s until the Japanese economy stagnated in the 1990s, while the positive cubic term produces a slight increase in the last years and may thus exaggerate how the CO_2 intensity falls at a diminishing rate when GDP per capita increases. The mid-1970s were a watershed in more than one sense. This was the time of the first energy crisis, but Japan was also at that time emerging from a period of rapid economic growth and industrialization implying possibly a rising CO_2 intensity after the late 1970s, which coincided with rapid economic growth and industrialization.

Lastly there are Brazil and India. In Brazil the CO_2 intensity has fluctuated without trend, and in India it rose until the early 1990s, but has fallen since. The model simulations shown in the diagrams for these countries explain very little or nothing of what has happened.

¹ Using a logarithmic model for China gives a much better fit and still produces diminishing rate of falling CO₂ intensity as GDP per capita rises.



Figure 2: Actual and simulated CO₂ intensity (kg per dollar GDP) in the six largest economies of the world.

The results for Japan, Brazil and India suggest that there may be a phase in the development of poor and medium rich countries where the CO_2 intensity of GDP increases with GDP per capita, in order to fuel rapid industrialization. Figure 3 shows the CO_2 intensity and the GDP per capita in two countries, Singapore and Thailand, that have experienced rapid economic growth. Singapore appears to have had a phase of increasing and then high CO_2 intensity during its first phase of rapid development up until about 1980. After that the CO_2 intensity has fallen rather evenly, but seems recently to have reached a plateau. In Thailand the CO_2 intensity grew with GDP per capita until 1997, but has since been fairly steady.



Figure 3: CO₂ intensity (left axis) and GDP per capita (right axis) in Singapore and Thailand.

A panel data approach

For the majority of countries, CO_2 intensity appears to fall as they get richer, and for these the relationship is non-linear in the majority of cases, implying that the CO_2 intensity falls at a diminishing rate. This is supported by estimating the second degree equation for the entire panel of data, with country-specific dummy variables. The results are shown in Table 1, with dummies omitted.

Table 1

Results from estimating the equation $y = a + b_1 x + b_2 x^2$, where y is CO₂ intensity and x is GDP per capita, with t-values in parentheses.

a	b_1	b_2	R ²
1.028136	0000147	7.23e-11	0.8244
(25.07)	(-13.49)	(6.54)	

The estimated curve is shown in Figure 4, together with the CO_2 intensity in selected countries, adjusted to the level of the United States, which is used as base for the dummies. The data for Thailand, the United States, the United Kingdom and Singapore were shown in Figures 2 and 3 and commented on in the previous section. Thailand and Singapore do not follow this overall tendency at all in their early phase. Data for the three richest countries in the world in 2014, Luxembourg, Norway and Switzerland, are also shown. The CO_2 intensity for the latter two is fairly flat. The CO_2 intensity for Luxembourg falls rapidly in the beginning, but is fairly flat in later years. Luxembourg is an example of a country that has developed rapidly towards a service-based, wealthy economy.

It could be argued that the results in Table 1 are biased because we have an unbalanced panel. For many countries data are not reported for the early years; there is a large influx of countries in the early 1990s, associated with the downfall of the Soviet Union and the disappearance of the iron curtain. Estimating the equation for data from 1992 onwards still gives significant coefficients with the same sign, but their numerical values now produce a U-shaped curve with a minimum at a GDP per capita of about 70,000 dollars. It is unlikely that the CO_2 intensity will begin to increase again at higher GDP levels, so we take this as a further evidence that the CO_2 intensity does indeed fall with GDP per capita, but at a diminishing rate.



Figure 4: The equation $y = a - b_1 x + b_2 x^2$ ($y = CO_2$ intensity, x = GDP per capita), as estimated for the entire panel of countries, and the CO₂ intensity of 7 selected countries.

3. World mineral production

One reason why world emissions of CO_2 falls at a declining rate is that the "dematerialization" of GDP is less easy than looking at the rich countries of the world might lead one to believe. It is well known that the share of minerals and industrial production in GDP (gross domestic product) declines as countries get richer (Radetzki and Wårell, 2017; Humphreys, 2013). Looking at the material content of GDP in rich countries may, however, exaggerate the possibilities of "dematerialization" of GDP for the world as a whole, because rich countries have outsourced their industrial production and even extraction of minerals to the so-called developing countries.²

How far can dematerialization of GDP be pushed? Some economists take a sanguine view:

"A central question of environmental policy is whether it will be possible to reach an absolute decoupling of economic growth measured in monetary terms at constant prices from the extraction of natural resources measured in physical units." (Meyer, Meyer and Distelkamp, 2012, p. 145).

 $^{^{2}}$ Helm (2012) discusses outsourcing of CO₂-intensive production to developing countries and shows how a decrease in CO₂ emissions for the UK turn into an increase if one accounts for emissions on basis of consumption rather than production.

Were this to happen, our material needs would have to be already satisfied, so that further economic growth would all be about immaterial things. In a world where the population is still growing and millions of people are still living in poverty, we seem to be far from such "absolute decoupling." In this section we look at the extraction of eleven minerals important for industrial production and compare it with the development of world GDP and population.

The data cover the period 1960-2017. Data on mineral extraction are from the US Geological Survey.³ The minerals we examine are bauxite, copper, cobalt, iron ore, lead, magnesium, molybdenum, nickel, platinum, tin and zinc. Of these, iron ore is probably the one most closely related to industrialization, while minerals such as magnesium, molybdenum, nickel and platinum are used in industrial processes and blended in with iron to produce steel of specific qualities. Bauxite is the raw material for aluminum, which we find in a variety of uses. We leave out precious minerals (gold and silver), which are not closely related to industrialization, and some minor minerals used in industrial processes and steel alloys (vanadium, for example).

Figure 5 compares the development in extraction of the aforementioned eleven minerals with the development of GDP and world population. There is limited support of the hypothesis that the world economy is becoming less dependent on minerals. The extraction of five minerals (bauxite, cobalt, nickel, molybdenum and platinum) has grown faster or at about the same rate as GDP from 1960 to 2017. All of these are used in industrial processes. Cobalt is used in batteries, so its increasing use is related to the development of the digital age. Extraction of four minerals (copper, iron ore, magnesium and zinc) has grown more slowly than GDP, but faster than world population. Extraction of tin and lead has grown at about the same rate as world population. In 2017 world population was 2.5 times larger than in 1960, while world GDP was more than seven times larger.

A conspicuous feature of Figure 5 is that the extraction of most minerals began to grow more quickly in the late 1990s or early 2000s. For cobalt and copper this began in the mid-1990s, and for bauxite, magnesium, molybdenum and zinc a little later (around 2000). It is tempting to see this as a result of the rapid economic growth in China and some other poor countries from the 1990s onwards. For iron ore we have particularly clear evidence. The production of iron ore took a sudden leap upwards 2002-2009, but has had a relatively stable growth after that. China's imports of iron ore rose from 70 million tonnes in 2000 to 620 in 2010, which coincided with a doubling of world production of iron ore (Hellmer and Ekstrand, 2013). As earlier discussed (see Figure 1), the stagnation of the CO₂ intensity of world GDP around 2000 is particularly noteworthy. China's industrialization has also left its indelible mark on time series of energy carriers (coal, oil, not discussed here) and emissions of carbon dioxide.

³ These refer to production from mines. An exception is cobalt, where there is serious disagreement between data for the periods 1960-1986 and 1983-2017. We have instead used data on production of refined metal. Data for this are missing for 1960-1968, but we have calculated these from mine production data, which are closely correlated with production of refined metal.





Figure 5: Production of eleven minerals, world population and world GDP. Index numbers, with 1960 at 100.

4. Conclusion

A falling CO_2 intensity as GDP per capita grows would contribute to reconciling economic growth and reduction in CO_2 emissions. But there is considerable evidence that this is primarily the case in rich countries and that the effect becomes smaller and smaller as countries get still richer. This will increase the burden on alternative technologies to deal with emissions. Furthermore, the need for alternative technologies will increase if the poor and medium rich countries of the world must go through a phase of increased energy use as they grow out of poverty. Hence, reconciling economic growth and reduction in CO_2 emissions would seem to depend critically on the development of energy sources other than fossil fuels. Economic growth by itself will not sweep this problem away.

The fall in CO_2 intensity as countries get richer may in fact exaggerate the effect of getting richer and developing a service economy. Parallel to this, rich countries are increasingly importing their energy intensive goods from poorer countries. The development in mineral extraction since 1960 shows few signs of dematerialization of world GDP. The production of nine out of eleven minerals important for industrial production has grown faster than world population, and five have grown faster than world GDP. Rather than decline, the growth of mineral production accelerated from the 1990s onwards, while in recent years growth has declined slightly. Mineral production is unlikely to become stabilized in the near future, provided that abatement of poverty in the world will continue. This is not going to make reductions in CO_2 emissions any easier.

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APPENDIX

Table A1

			y = a + l	<i>bx</i>	y = 0	$a + bx + cx^2$			y = a + bx +	$cx^2 + dx^3$	
	GDPcapita	Period	b	\mathbb{R}^2	b	С	\mathbb{R}^2	b	С	d	\mathbb{R}^2
Luxembourg	107152.9	1960-	0000124***	0.8037	0000494***	2.72e-10***	0.9462				
_		2014	(14.73)		(15.53)	(11.74)					
Norway	89274.96	1960-	-1.25e-06***	0.5916	1.39e-06	-2.25e-11**	0.6491	.0000245***	-4.61e-10***	2.54e-15***	0.8285
		2014	(8.76)		(1.52)	(2.92)		(7.59)	(7.65)	(7.30)	
Bermuda	79251.78	1960-	-3.72e-07	0.0708							
		2013	(1.99)								
Switzerland	76410.86	1980-	-2.30e-06***	0.9438	-3.59e-07	-1.48e-11	0.9453				
		2014	(23.53)		(0.17)	(0.94)					
Macao	69749.16	1982-	-2.52e-06***	0.8655	-3.91e-06**	1.63e-11	0.8725				
		2014	(14.12)		(3.59)	(1.29)					
Qatar	67901.22	2000-	0000433***	0.8120	0001377	7.17e-10	0.8147				
		2014	(7.49)		(0.61)	(0.42)					
Denmark	59437.93	1960-	-6.67e-06***	0.8632	2.38e-06	-1.07e-10**	0.8864	.0000532**	-1.40e-09***	1.04e-14**	0.9084
		2014	(18.28)		(0.85)	(3.26)		(3.61)	(3.78)	(3.50)	
Australia	54546.2	1960-	-4.67e-06***	0.7760	.0000101***	-2.00e-10***	0.9045	.0000347**	-9.16e-10**	6.59e-15*	0.9159
		2014	(13.55)		(5.67)	(8.37)		(3.65)	(3.35)	(2.63)	
Ireland	54052.95	1970-	-8.20e-06***	0.9322	0000179***	1.47e-10***	0.9543				
		2014	(24.31)		(8.23)	(4.50)					
Sweden	53561.89	1960-	-9.81e-06***	0.8342	0000197***	1.34e-10*	0.8495				
		2014	(16.33)		(4.52)	(2.30)					
Singapore	52244.44	1960-	0000143***	0.3515	.000011	-5.00e-10*	0.4231	.0001354***	-6.32e-09***	7.34e-14***	0.6707
		2014	(5.36)		(1.07)	(2.54)		(6.28)	(6.64)	(6.19)	
United States	50871.67	1960-	0000184***	0.9532	0000299***	1.66e-10**	0.9591				
		2014	(32.85)		(7.10)	(2.75)					
Netherlands	50497.24	1960-	-7.11e-06***	0.8496	-5.47e-07	-9.36e-11*	0.8637				
		2014	(17.31)		(0.19)	(2.32)					
Canada	50221.84	1960-	0000107***	0.8902	-1.29e-06	-1.36e-10*	0.9022				
		2014	(20.73)		(0.34)	(2.53)					

Countries in the sample, their GDP per capita (2010 US dollars in 2014), results for linear model (y = a + bx), a model with a squared ($y = a + bx + cx^2$) and a cubic term ($y = a + bx + cx^2 + dx^3$), where y is CO₂ intensity, x is GDP per capita, and with t-values in parentheses.

Austria	47922.34	1960-	-5.65e-06***	0.9200	-7.72e-06***	3.29e-11	0.9230				
		2014	(24.69)		(5.23)	(1.42)					
Japan	46484.16	1960-	-4.05e-06***	0.6848	3.99e-06	-1.39e-10***	0.7539	.0000428***	-1.71e-09***	1.89e-14***	0.8733
-		2014	(10.73)		(1.87)	(3.82)		(7.37)	(7.50)	(6.93)	
Greenland	46443.76	1970-	0000102***	0.5001	0000473***	5.80e-10**	0.6208				
		2014	(6.56)		(4.62)	(3.66)					
Finland	45239.37	1960-	-5.45e-06***	0.4797	.0000121**	-2.85e-10***	0.6114	.0001114***	-3.84e-09***	3.90e-14***	0.7890
		2014	(6.99)		(2.86)	(4.20)		(7.20)	(7.05)	(6.55)	
Germany	45022.57	1991-	0000113***	0.9445	0000556***	5.67e-10***	0.9694				
		2014	(19.34)		(5.19)	(4.14)					
Iceland	44775.64	1996-	-7.71e-06***	0.8477	0000339**	3.33e-10	0.8749				
		2014	(9.73)		(2.41)	(1.86)					
Belgium	44676.66	1960-	0000175***	0.9427	0000359***	3.04e-10***	0.9646				
		2014	(29.52)		(10.99)	(5.68)					
France	41374.76	1960-	0000129***	0.9395	0000141***	2.15e-11	0.9397				
		2014	(28.69)		(4.28)	(0.37)					
United	40908.75	1960-	0000208***	0.9221	0000644***	7.83e-10***	0.9862				
Kingdom		2014	(25.05)		(22.75)	(15.51)					
Andorra	40785.05	1990-	-3.95e-06***	0.5800	0000108	8.33e-11	0.5871				
		2014	(5.64)		(0.97)	(0.61)					
United Arab	39146.11	1975-	-1.65e-06**	0.1643	-6.92e-06	3.58e-11	0.2047				
Emirates		2014	(2.73)		(1.78)	(1.37)					
Kuwait	36259.67	1995-	-9.40e-06***	0.5362	.0000645	-8.73e-10	0.6170	.0001131	-2.04e-09	9.29e-15	0.6171
		2014	(4.56)		(1.65)	(1.89)		(0.14)	(0.10)	(0.06)	
New Zealand	36142.52	1970-	-2.96e-06**	0.2218	.0000547***	-1.02e-09***	0.5744	-7.57e-06	1.25e-09	-2.71e-14	0.5792
		2014	(3.50)		(5.59)	(5.90)		(0.08)	(0.37)	(0.68)	
Hong Kong	35717.68	1961-	-4.98e-06***	0.7657	-3.51e-08	-1.35e-10**	0.8077				
		2014	(13.04)		(0.02)	(3.34)					
Italy	33615.97	1960-	-3.27e-06***	0.4271	.0000133***	-3.28e-10***	0.6269	.0001052***	-4.28e-09***	5.29e-14***	0.8744
		2014	(6.29)		(4.20)	(5.28)		(11.25)	(10.81)	(10.03)	
Brunei	33313.83	1974-	6.21e-06	0.0455							
		2014	(1.36)								
Israel	32661.29	1960-	-3.11e-06***	0.2639	1.72e-06	-1.16e-10	0.2792	0000637**	3.30e-09**	-5.53e-14**	0.4142
		2014	(4.36)		(0.37)	(1.05)		(3.26)	(3.29)	(3.43)	
Spain	29496.38	1960-	-9.57e-07	0.0317							
		2014	(1.32)								
Bahamas	27246.48	1960-	0000985***	0.3828	0002906	3.87e-09	0.3938				
		2014	(5.73)		(1.46)	(0.97)					

S Korea	24323.57	1960-	0000154***	0.6618	-7.51e-06	-3.46e-10	0.6729				
		2014	(10.18)		(1.22)	(1.33)					
Malta	23676.03	1970-	0000133***	0.5576	0000125	-3.16e-11	0.5577				
		2014	(7.36)		(1.26)	(0.09)					
Slovenia	23224.4	1995-	000019***	0.8604	000073*	1.34e-09*	0.8902				
		2014	(10.53)		(2.90)	(2.15)					
Greece	22565.68	1960-	7.43e-06***	0.4673							
		2014	(6.82)								
Bahrain	22390.68	1980-	0000564***	0.4144	0010144**	2.41e-08**	0.5602				
		2014	(4.83)		(3.45)	(3.26)					
Portugal	21533.49	1960-	3.05e-06***	0.3580							
		2014	(5.44)								
Saudi Arabia	21183.46	1968-	0000196***	0.4650	.0000307	-9.33e-10	0.4922	.0005384**	-2.16e-08**	2.67e-13**	0.6103
		2014	(6.25)		(0.93)	(1.54)		(3.75)	(3.76)	(3.61)	
Czech Republic	20343.68	1992-	0000606***	0.9353	0001912**	3.93e-09*	0.9491				
		2014	(17.42)		(3.41)	(2.33)					
Cyprus	20009.06	1975-	0000111***	0.7746	0000263***	4.91e-10*	0.8052				
		2014	(11.43)		(4.14)	(2.41)					
Slovak Republic	18003.54	1992-	0000619***	0.9311	0002066***	5.59e-09***	0.9764				
		2014	(16.85)		(8.81)	(6.19)					
Estonia	17453.37	1995-	0000718***	0.7729	0003915***	1.26e-08***	0.9265				
		2014	(7.83)		(7.26)	(5.96)					
Oman	17167.05	1965-	.0000358***	0.4026							
		2014	(5.69)								
Trinidad and	16641.74	1960-	.0000679***	0.2140							
Tobago		2014	(3.80)								
Equatorial	16028.25	1980-	-8.85e-07	0.0015							
Guinea		2014	(0.824)								
Barbados	15901.9	1974-	.000023***	0.5875							
		2014	(7.45)								
St. Kitts and	15029.62	1977-	9.97e-06***	0.5907							
Nevis		2014	(7.21)								
Lithuania	14935.54	1995-	000046***	0.8058	0001993***	7.68e-09***	0.9239				
		2014	(8.64)		(6.63)	(5.14)					
Chile	14681.33	1960-	0000171***	0.5044	0000515**	2.00e-09*	0.5518				
		2014	(7.35)		(3.47)	(2.34)					
Hungary	14119.07	1991-	0000631***	0.9369	0001514*	3.93e-09	0.9421				
		2014	(18.08)		(2.34)	(1.36)					

Poland	14090.62	1990-	0001256***	0.8801	0004851***	1.86e-08***	0.9813				
		2014	(13.00)		(14.62)	(10.91)					
Uruguay	13856.7	1960-	0000218***	0.4520	00011***	4.95e-09***	0.5994				
		2014	(6.61)		(5.40)	(4.37)					
Latvia	13758.96	1995-	000041***	0.7653	000205***	8.59e-09***	0.9041				
		2014	(7.66)		(6.17)	(4.96)					
Venezuela	13709.04	1960-	0000607***	0.7334	0001989*	5.32e-09	0.7484				
		2014	(12.08)		(2.53)	(1.76)					
Croatia	13651.99	1995-	00002***	0.6218	.0000528	-3.09e-09	0.6538	.0008233	-7.02e-08	1.92e-12	0.7043
		2014	(5.44)		(0.91)	(1.25)		(1.75)	(1.72)	(1.65)	
Turkey	13312.46	1960-	.0000134***	0.3570							
		2014	(5.43)								
Seychelles	12850.49	1963-	.0000613***	0.5217							
-		2014	(7.39)								
Antigua and	12403.53	1977-	0000508***	0.2989	0002597**	1.11e-08**	0.4426				
Barbuda		2014	(3.92)		(3.68)	(3.00)					
Brazil	11870.15	1960-	1.41e-07	0.0006							
		2014	(0.18)								
Russia	11865.03	1992-	0001394***	0.8973	0003123*	9.97e-09	0.9077				
		2014	(13.55)		(2.70)	(1.50)					
Kazakhstan	10646.03	1992-	0001602***	0.4743	0005437	2.77e-08	0.5126				
		2014	(4.35)		(1.77)	(1.25)					
Malaysia	10398.23	1970-	.0000282***	0.5477							
		2014	(7.22)								
Panama	10350.4	1960-	0000234***	0.2020	0000474*	2.00e-09	0.3503				
		2014	(5.17)		(2.09)	(1.08)					
Argentina	10323.21	1960-	-7.82e-06*	0.0840	.000101**	-6.71e-09**	0.2506	.0006925*	-8.25e-08*	3.16e-12*	0.3071
		2014	(2.20)		(3.14)	(3.40)		(2.37)	(2.22)	(2.04)	
Palau	9692.272	2000-	0000805***	0.6444	0004123	1.69e-08	0.6533				
		2014	(4.85)		(0.69)	(0.55)					
Mexico	9536.6	1960-	.0000125***	0.2135							
		2014	(3.79)								
Gabon	9508.285	1960-	.0000433***	0.3581							
		2014	(5.44)								
Romania	9227.437	1990-	0001598***	0.7872	0004402	2.09e-08	0.8008				
		2014	(9.22)		(1.91)	(1.22)					
Caribbean small	9169.713	1966-	.000038*	0.1170							
states		2014	(2.50)								

Mauritius	9163.633	1976-	.0000236***	0.5484					
		2014	(6.70)						
Costa Rica	9065.026	1960-	8.90e-06***	0.2141					
		2014	(3.80)						
Suriname	8942.961	1975-	0000894***	0.4108	0003837	2.06e-08	0.4296		
		2014	(5.15)		(1.44)	(1.11)			
St. Lucia	8147.524	1977-	.0000126**	0.2183					
		2014	(3.17)						
Maldives	8124.708	1995-	.0000445***	0.6354					
		2014	(5.60)						
Grenada	7932.668	1977-	.0000297***	0.6476					
		2014	(8.13)						
South Africa	7582.553	1960-	0000259	0.0133					
		2014	(0.85)						
Botswana	7574.282	1972-	5.10e-06	0.0061					
		2014	(0.50)						
Lebanon	7447.364	1988-	0000483**	0.3100	0000759	2.19e-09	0.3124		
		2014	(3.35)		(0.79)	(0.77)			
Bulgaria	7299.549	1980-	0004044***	0.6383	0017293*	1.25e-07*	0.6821		
		2014	(7.63)		(2.73)	(2.10)			
Colombia	7291.692	1960-	0000615***	0.8838	0000944***	3.69e-09	0.8910		
		2014	(20.07)		(5.24)	(1.85)			
Montenegro	7045.116	2005-	0000336	0.0589					
		2014	(0.71)						
Dominica	6951.032	1977-	.0000289***	0.7175					
		2014	(9.56)						
Libya	6697.103	1999-	0000908***	0.9038	0001665**	4.32e-09	0.9175		
		2014	(11.47)		(3.19)	(1.47)			
Belarus	6664.097	1992-	0003874***	0.7882	0014156***	1.18e-07**	0.8712		
		2014	(8.84)		(4.91)	(3.59)			
St. Vincent and	6467.158	1960-	.0000455***	0.7939					
the Grenadines		2014	(14.29)						
Turkmenistan	6399.271	1992-	0004186***	0.8165	0006845*	3.36e-08	0.8235		
		2014	(9.67)		(2.27)	(0.89)			
Dominican	6203.726	1960-	.0000202	0.0669					
Republic		2014	(1.95)						
Cuba	6182.774	1970-	0001385***	0.6700	0002693*	1.56e-08	0.6801		
		2014	(9.34)		(2.35)	(1.15)			

Iran	6161.104	1960-	0000387	0.0477							
		2014	(1.63)								
Azerbaijan	6122.98	1992-	0005284***	0.7735	0015485**	1.40e-07*	0.8211				
		2014	(8.47)		(3.47)	(2.31)					
China	6108.239	1960-	0006465***	0.5966	0019432***	2.50e-07***	0.8094				
		2014	(8.85)		(10.95)	(7.62)					
Namibia	5901.243	1990-	.0000148	0.0405							
		2014	(0.335)								
Peru	5825.198	1960-	3.13e-06	0.0066							
		2014	(0.59)								
Serbia	5593.061	2006-	0005605**	0.6879	0004153	-1.38e-08	0.6879				
		2014	(3.93)		(0.06)	(0.02)					
Thailand	5591.106	1960-	.0000999***	0.7690							
		2014	(13.28)								
Ecuador	5428.714	1960-	.0001344***	0.6067							
		2014	(9.04)								
Iraq	5253.627	1968-	000293***	0.7105	0008558***	8.76e-08***	0.7851				
		2014	(10.51)		(5.86)	(3.91)					
Bosnia and	4992.949	1994-	.0000665**	0.3564							
Herzegovina		2014	(3.24)								
Macedonia	4920.216	1992-	0005975***	0.9559	0003798	-2.78e-08	0.9563				
		2014	(21.34)		(0.69)	(0.40)					
Jamaica	4714.861	1966-	.0000113	0.0016							
		2014	(0.27)								
Algeria	4675.885	1960-	.0002051***	0.4197							
_		2014	(6.19)								
Albania	4413.562	1980-	0001537*	0.1574	.0004372	-1.01e-07	0.1998	.0046676*	-1.73e-06*	1.94e-10*	0.3149
		2014	(2.48)		(0.95)	(1.30)		(2.45)	(2.41)	(2.28)	
Belize	4411.856	1960-	0001038***	0.6063	.000061	-2.95e-08*	0.6398	.0005138	-2.11e-07	2.20e-11	0.6542
		2014	(9.03)		(0.81)	(2.20)		(1.60)	(1.68)	(1.45)	
Tunisia	4271.327	1965-	-3.19e-06	0.0017							
		2014	(0.28)								
Fiji	4084.2	1960-	0000159	0.0230							
· ·		2014	(1.12)								
Swaziland	3980.774	1970-	0001052***	0.4098	0006246***	1.02e-07***	0.5918				
		2014	(5.46)		(5.16)	(4.33)					
Mongolia	3901.867	1981-	0000532	0.0043							
-		2014	(0.37)								

Georgia	3851.723	1992-	0001869	0.1578							
		2014	(1.98)								
Armenia	3827.343	1992-	0001845***	0.4781	0008747**	1.49e-07*	0.6036				
		2014	(4.39)		(3.16)	(2.52)					
Paraguay	3761.912	1960-	.0000304***	0.3593							
		2014	(5.45)								
Angola	3746.66	1980-	.0000233	0.0402							
		2014	(1.17)								
Indonesia	3692.943	1960-	.0000882***	0.6058							
		2014	(9.03)								
Guyana	3595.925	1960-	0001406**	0.1803	.0004233	-1.17e-07	0.2272	.0072403**	-3.02e-06**	3.94e-10**	0.3766
		2014	(3.41)		(1.32)	(1.78)		(3.67)	(3.63)	(3.50)	
Tonga	3581.837	1981-	.0000675***	0.3580							
		2014	(4.22)								
Samoa	3524.596	1982-	0000188***	0.3414	0001796	2.75e-08	0.4058				
		2014	(4.01)		(2.01)	(1.80)					
Sri Lanka	3506.871	1961-	0000411**	0.2043	0002013***	4.37e-08**	0.3578				
		2014	(3.65)		(4.28)	(3.49)					
Cabo Verde	3369.643	1980-	.0000132	0.0278							
		2014	(0.97)								
Jordan	3348.827	1975-	-6.72e-06	0.0004							
		2014	(0.13)								
Marshall Islands	3333.361	1990-	.0001477*	0.2084							
		2014	(2.46)								
El Salvador	3272.74	1965-	.0000945*	0.1275							
		2014	(2.65)								
Tuvalu	3196.979	1990-	0000265	0.0262							
		2014	(0.79)								
Morocco	3160.526	1966-	.0000594***	0.3728							
		2014	(5.29)								
Pacific island	3116.11	1981-	0000742**	0.1964	.0001736	-4.62e-08	0.1997	.009214	-3.40e-06	4.14e-10	0.2120
small states		2014	(2.80)		(0.25)	(0.36)		(0.69)	(0.69)	(0.68)	
Guatemala	3007.9	1960-	.0000865***	0.5092							
		2014	(7.41)								
Ukraine	2967.213	1992-	001246***	0.6467	0034351	4.43e-07	0.6580				
		2014	(6.20)		(1.28)	(0.82)					
Congo, Rep.	2922.973	1960-	0000162	0.0092							
		2014	(0.70)								

Vanuatu	2909.775	1979-	0000567	0.0882					
		2014	(1.81)						
Micronesia	2716.323	1992-	.0001346	0.0831					
		2014	(1.38)						
Egypt	2608.375	1960-	.0000182	0.0184					
		2014	(1.00)						
Nigeria	2563.092	1960-	-5.34e-06	0.0002					
		2014	(0.1)						
Timor-Leste	2547.159	2002-	0000294**	0.6145	000018	-2.14e-09	0.6161		
		2014	(4.19)		(0.32)	(0.20)			
West Bank and	2529.996	1997-	0000256	0.0090					
Gaza		2014	(0.37)						
Philippines	2505.819	1960-	-2.22e-06	0.0001					
		2014	(0.07)						
Bhutan	2500.26	1980-	.0000891*	0.1410					
		2014	(2.33)						
Papua New	2329.891	1960-	.0001971**	0.2020					
Guinea		2014	(3.66)						
Bolivia	2317.257	1960-	.0004785***	0.3062					
		2014	(4.84)						
Honduras	2059.475	1960-	.000298***	0.5725					
		2014	(8.42)						
Moldova	1986.941	1996-	0011563**	0.3379	0040152	1.01e-06	0.3571		
		2014	(3.03)		(1.00)	(0.71)			
Sudan	1837.138	1960-	0000578	0.0338					
		2014	(1.36)						
Nicaragua	1812.995	1960-	0002158***	0.4642	0006532*	1.25e-07	0.4885		
		2014	(6.78)		(2.33)	(1.57)			
Uzbekistan	1744.491	1992-	0049982***	0.9556	012044***	2.99e-06***	0.9777		
		2014	(21.27)		(7.57)	(4.45)			
Ghana	1659.797	1960-	0000489	0.0276					
		2014	(1.23)						
India	1645.326	1960-	.0000117	0.0009					
		2014	(0.22)						
Zambia	1620.823	1964-	.0004173***	0.3429					
		2014	(5.06)						
Kiribati	1565.243	1970-	0000653***	0.2867	000298**	4.22e-08*	0.3880		
		2014	(4.16)		(3.33)	(2.64)			

Vietnam	1565.02	1984-	.0004117***	0.6159							
		2014	(6.82)								
Solomon Islands	1475.528	1990-	0001713***	0.6285	-2.37e-06	-6.23e-08	0.6306				
		2014	(6.24)		(0.00)	(0.35)					
Laos	1470.5	1984-	.0000906*	0.1909							
		2014	(2.62)								
Cameroon	1428.216	1960-	.0002393***	0.3034							
		2014	(4.80)								
Cote d'Ivoire	1384.91	1960-	1.37e-06	0.0000							
		2014	(0.04)								
Mauritania	1326.159	1960-	.0000829	0.0011							
		2014	(0.24)								
Lesotho	1323.238	1990-	000699***	0.9588	0022598***	7.70e-07**	0.9953				
		2014	(23.13)		(18.77)	(13.01)					
Myanmar	1266.124	1960-	0006129***	0.6764	0016077***	8.07e-07***	0.7516				
		2014	(10.53)		(6.28)	(3.97)					
Sao Tome and	1241.459	2001-	.0000666	0.0793							
Principe		2014	(1.02)								
Pakistan	1111.196	1960-	0000218	0.0021							
		2014	(0.34)								
Yemen	1101.117	1990-	.0004162**	0.2800							
		2014	(2.99)								
Kenya	1075.659	1960-	0005825***	0.6512	0017443**	7.69e-07*	0.6800				
		2014	(9.95)		(3.23)	(2.17)					
Senegal	1018.393	1960-	0008218***	0.2097	.0084957	-4.98e-06	0.2553	0778848	.0000877	-3.30e-08	0.2693
_		2014	(3.75)		(1.63)	(1.78)		(0.89)	(0.93)	(0.99)	
Kyrgyzstan	1003.51	1992-	0005907	0.0338							
		2014	(0.86)								
Cambodia	972.9792	1993-	.0000207	0.0082							
		2014	(0.41)								
Chad	967.1028	1960-	0000398	0.0644							
		2014	(1.91)								
Zimbabwe	939.7803	1964-	.0002865	0.0707							
		2014	(1.93)								
Bangladesh	922.1611	1972-	.0005531***	0.7332					1		
Ŭ		2014	(10.61)								
Tajikistan	892.64	1992-	0007048	0.1292							
5		2014	(1.77)								
			1 . /								

Benin	833.6409	1960-	.0022878***	0.8609							
		2014	(18.11)								
Tanzania	782.6772	1988-	.0002226***	0.4529							
		2014	(4.55)								
Comoros	779.8398	1980-	0006535***	0.5358	.0042805	-2.93e-06	0.5584	.2189843*	0002595*	1.02e-07*	0.6357
		2014	(6.17)		(1.11)	(1.28)		(2.61)	(2.59)	(2.57)	
Haiti	728.7803	1996-	0015115***	0.6652	.030579**	0000222**	0.7919	.2185652	0002844	1.22e-07	0.7599
		2014	(5.81)		(2.97)	(3.12)		(0.60)	(0.56)	(0.52)	
Guinea	714.1633	1986-	.0001323	0.0781							
		2014	(1.51								
Mali	705.7885	1967-	0000382	0.0774							
		2014	(1.97)								
Nepal	675.7353	1960-	.000779***	0.8416							
		2014	(16.78)								
Rwanda	672.6396	1960-	0000623	0.0056							
		2014	(0.54)								
Uganda	642.8774	1982-	.0001271***	0.3780							
-		2014	(4.34)								
Burkina Faso	639.7096	1960-	.0003926***	0.5212							
		2014	(7.60)								
Sierra Leone	562.8597	1960-	.0003473	0.0353							
		2014	(1.39)								
Guinea-Bissau	545.8985	1970-	0000317	0.0014							
		2014	(0.24)								
Togo	531.1561	1960-	000219	0.0058							
_		2014	(0.55)								
Gambia	530.3189	1966-	.0018157***	0.7183							
		2014	(10.95)								
Eritrea	514.1796	1994-	.0006765	0.0856							
		2011	(1.22)								
Mozambique	493.2533	1980-	0012262**	0.2167	0071831**	.00001*	0.3306				
		2014	(3.02)		(2.78)	(2.33)					
Malawi	484.3686	1964-	0007698***	0.4745	0010358	3.58e-07	0.4750				
		2014	(6.65)		(0.89)	(0.23)					
Ethiopia	452.7782	1981-	0001569	0.0754							
		2014	(1.62)								
Madagascar	408.661	1960-	0002075***	0.2204	0009911	6.96e-07	0.2338				
_		2014	(3.87)		(1.20)	(0.95)					

Congo, Dem.	397.582	1960-	.0000631***	0.4048							
Rep.		2014	(6.00)								
Niger	386.7258	1960-	0005727***	0.5421	.0014242	-2.00e-06**	0.6030	.0238315***	0000478***	3.01e-08***	0.7331
		2014	(7.92)		(2.01)	(9.82)		(5.26)	(5.19)	(4.99)	
Liberia	376.5889	1960-	0001825**	0.1628	0022406***	1.17e-06***	0.6997				
		2014	(3.21)		(10.36)	(9.64)					
Central African	302.5465	1960-	0002818***	0.4184	0011345*	8.70e-07	0.4480				
Republic		2014	(6.18)		(2.21)	(1.67)					
Burundi	243.1019	1962-	.000259*	0.0907							
		2014	(2.26)								