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## The Structure and Dynamics of International Development Assistance

**Abstract:** We study the structure of international aid coordination by creating and analyzing a tripartite network of donor organizations, recipient countries and development issues using web-based information. We develop a measure of coordination and find that it is moderate, achieving about 60% of its theoretical maximum. Many countries are strongly connected to organizations that are related to the issues that are salient there. Nevertheless, we identify many countries that are poorly served, issues that are inadequately attended to, and organizations that focus on the wrong combination of places and issues. Our approach may be used to improve decentralized coordination.

**Keywords:** aid coordination; development assistance; network science.

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## 1 Introduction

This paper explores the structure of international development assistance by creating a tri-partite network composed of aid organizations, countries of interest and development issues. The network was constructed using online data on the co-occurrence of pairs of organizations, countries and issues in the websites of development agencies and in the World Wide Web. We compare the structure of networks connecting countries to issues, countries to organizations and organizations to issues, to assess the degree to which the system is able to match a country with organizations that have expertise in the issues that are salient in that country. More generally, this paper extends the use of data-mining tools and network science to the study of decentralized networks of organizations, which is an area that hitherto eluded empirical analysis.

International development assistance became an explicit policy issue in the 1940s. It arose from two fundamental changes to the global economy: the Great Divergence (Pritchett 1997) and the proliferation of states. Over the past two centuries, incomes across countries and regions of the world have greatly diverged, from a ratio of about 4 to 1 in 1820 to a ratio of over 100 to 1 at present (Maddison 2008). Initially, these rising gaps occurred under colonialism. Income gaps grew within large multi-ethnic empires, such as that of the British or the French. It was then that Rudyard Kipling (1899) wrote about the *White Man's Burden*, based on the idea that it was the responsibility of Western empires to raise the level of development of their poorer possessions. In France, the organizing principle was “*Le Rayonnement de la France*,” the illumination of the Asian, African and Caribbean possessions of the French Empire from its culturally-sunny and ethnically French core.

After World War II, however, the decolonization process led to the dismemberment of the British and French empires and a concomitantly rapid increase in the number of new sovereign states. The large and widening income gaps now occurred between, rather than within, sovereign nations. The term “international” in “international development,” therefore, highlighted the fact that it was no longer about sovereigns managing their possessions, but about rich ones trying to help poorer ones catch up.

One part of the international development effort was organized multilaterally through either the United Nations or the Bretton Woods organizations (i.e., the IBRD, a.k.a. the World Bank, the IMF and the GATT<sup>1</sup>). Simultaneously, rich countries developed their own international development efforts, which lead to bilateral organizations such as USAID (for the US), DFID (UK), CIDA (Canada), SIDA (Sweden), where the acronym ID stands for International Development and the letter A for Agency. Other nations preferred to use the term “cooperation” instead of “development” for their bilateral efforts. These include the Japanese, Koreans, Spaniards and the Dutch.

In addition to official multilateral and bilateral organizations, private philanthropic organizations had emerged from Church groups (e.g., CARE), wealthy individuals (e.g., Rockefeller, Ford, Gates), action-oriented groups (e.g., Medecins Sans Frontier, ACCION International) and name-and-shame organizations (e.g., Human Rights Watch). Moreover, starting around 1960, in different continents, regional multilateral development banks were formed, modeled on the World Bank. This was the case of the AfDB, ADB and IADB, and more recently the EBRD. As a consequence, during the last 60 years the number of organizational actors in

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<sup>1</sup> For simplicity, we name organizations through their well-known acronyms. A list of acronyms and the organizational names is included in Table 1.

international development has greatly expanded, giving rise to a very diverse set of organizations that now need to coordinate their efforts.

The expansion, however, has not been limited only to the number of actors, but has also included the content of the issues at hand. Initially, the intellectual interpretation of the obstacles to development focused on the availability of capital and infrastructure. The successful experience of the Marshall Plan in Europe (1948–1952) suggested that this was an appropriate approach. The application of the same strategy through the IBRD in other parts of the world, however, led to disappointment, and the revised strategies identified a widening list of potential obstacles to development, including issues such as education, public health, demographic pressures, industrial bottlenecks, macroeconomic stability, appropriate market regulation, environmental sustainability, human rights, institutional quality, governance, inequality and gender gaps, among many others. As a consequence, different specialized topics were recognized as important and were buttressed by dedicated organizations and programs. The UN system responded by creating specialized bodies for different purposes such as the WHO (for health), UNESCO (education), UNIDO (industrial development), FAO (for agriculture and food security), UNDP (technical assistance for development), UNCTAD (trade), ILO (labor), UNHCR (human rights and refugees), etc. Concomitantly, the regional organizations created similar structures (e.g., the Organization of American States, the Pan-American Health Organization, ASEAN, Organization for African Unity).

Within the developed countries, governments started to involve an increasing number of ministries and departments, beyond their development agencies, in international cooperation efforts, in order to make available their areas of expertise, such as financial regulation, tax administration and health. For example, in the USA, some 14 different departments run development assistance programs. Within developing countries, a tendency towards decentralization of administration and authority led to more independent state and local governments with responsibility for many of the development projects.

Hence, economic divergence, decolonization, decentralization and the changing conceptual framework resulted in an increasing number of donors, recipients and issues, making international development a truly complex undertaking. The world now has hundreds of official aid organizations and thousands of private organizations working in over 140 sovereign states and many more sub-national governments. According to the Directory of Development Organization there are more than 72,000 related organizations around the world.<sup>2</sup> The multiplication of donors, recipients and issues implies a network with hundreds of millions of possible connections.

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<sup>2</sup> <http://www.devdir.org/stats.htm>, May 30th, 2012.

With the emergence of this increasingly complex structure, questions of coordination rose to the fore. How could donor organizations know which recipient organizations to work with and on what issues? How could donor organizations be sure that they were creating value rather than just duplicating the efforts of others? How could organizations be accountable for results if so many other players were involved in the process? How should resources be allocated across countries and topics? How could resources be allocated to the most effective organizations within each topic? How could political parties in donor countries support the use of public resources for international development if the beneficiaries could not be held accountable for results?

Given these challenges, the issue of aid coordination has become central to the whole development effort. One such coordinating tool is the convening of so-called consultative groups or roundtables. A Google search of the term “consultative group,” for instance, finds some 3.4 million pages. The conjunction of the terms “roundtable” + “aid” + “development” generates more than 10 million pages, while the term “aid coordination” generates more than 420,000 pages.

The purpose of this paper is to map the network of international aid efforts, quantify its structure and assess its performance. To do this, we use techniques that exploit the footprints left by the aid efforts on the web. We find that the aid community is held together by large, diversified organizations, which act as hubs that connect to smaller more specialized entities. This defines a more horizontal structure with large hubs where the distance between organizations is smaller than in a hierarchy. Also, aid organizations tend to organize their efforts geographically, except when dealing with the very large developing countries. Finally, aid organizations tend to care about the issues that are important for the countries they care about, although important misalignments are present, as we illustrate below.

## 2 Centralized vs. Decentralized Coordination

Centralized coordination requires hierarchies. Decentralized coordination lead to networks that are more horizontal. Decentralized coordination is not a new problem in economics. A founding idea of economics is the invisible hand: the notion of a decentralized process where agents following their own goals lead to systemic self-organization. Much of economics is focused on the study of the potential inefficiencies of self-organizing market coordination (e.g., missing markets, non-convexities, imperfect information, externalities, etc.) and of the policies that may facilitate improved outcomes. The discussion of economic

coordination reappeared after the Bolshevik Revolution, with questions regarding the viability of central planning. In the view of von Mises and Hayek (Hayek 1929, 1945), the issue hinged on an information problem. Markets can be seen as computing devices that can assess whether a particular good has a social value worth more than its inputs, and hence, worthwhile to be produced. In this view, markets mobilize information regarding relative preferences, subjective valuations and alternative production techniques that is decentralized among potentially many producers and consumers. By contrast, a central planner has no alternative process to gather and process this information and hence cannot do a decent coordination job. While some formal models were developed to prove the potential efficiency of central planning, with a fixed set of goods and technologies (e.g., Lange 1949; Kornai and Liptak 1965), the literature died out after it became clear that the central planner could hardly explore a set of expanding technological possibilities in the way that Schumpeterian entrepreneurs could. This became patently obvious in the comparison of product diversity and technological progress between the East and West during the Cold War (Kornai 1992).

Aid coordination faces a similar dilemma to that of markets: the choice between centralized control vs. self-organization. The coordination problem consists in the fact that there are many potential sources of supply and demand for a large set of alternative development goods. Aid organizations have goals that may or may not be altruistic, recipient agencies may be advancing national or particular interests and the transaction may involve a grant, a loan or technical assistance. The details of the transaction are less important than the fact that, whatever are the goals of each organization, only certain types of transactions will take place that are agreeable to both parties and their “authorizing environments” (Moore 1997). One can imagine a situation in which some central entity, at the global or national level, controls the process and allocates roles and tasks. For example, the current *de jure* approach to aid coordination, as framed by the Paris Declaration for Aid Effectiveness (2005) and the Accra Agenda for Action (2008) starts with an international agreement on the substantive targets to be achieved, namely, the millennium development goals (MDGs). Given these objectives, countries must then write a Poverty Reduction Strategy Paper, which is supposed to serve as a mechanism for donor coordination. This paper should be reviewed by the different domestic constituencies and be consistent with the achievement of the agreed MDGs. Once the paper is ready, roundtables and consultative groups are convened to allocate roles and functions to the different development partners.

In parallel to these formalized *de jure* processes there are the *de facto* efforts of each donor and each recipient to explore areas of common interest and commit to individual projects without reference to some over-arching

planning process. This is particularly so, given that recipient countries are composed of many individual and institutional players, each with their own agenda. This self-organizing process is based on the fact that each donor, recipient or service provider has significant autonomy and hence would only enter into voluntary agreements based on identified mutual goals, whatever these goals might be.

What the market has taught us is that self-organizing systems require information and feedback loops. In the market, the price system is the key source of information that allows decentralized coordination, as profit-motivated suppliers and utility-motivated consumers respond in opposite ways to price movements. In the case of international development, however, it is not clear what the information structure is and what brings potential supply and demand to some balance. The lack of an appropriate information structure may lead many participants to demand more explicit coordination.

One of the major differences between centralized vs. decentralized coordination is that the former requires a pre-definition of the preferences, while in the latter preferences are revealed through the self-organizing process itself. For example, the MDGs are an example of the centralized approach that sets targets *ex ante* on some development goals but not others. There are MDGs for education and health, but none for better jobs for the non-poor, crime reduction or urban transport, although these issues figure prominently in the preferences of voters in many developing countries. A self-organizing system would let goals be determined by the preferences of donors and recipients, according to their own priorities and views, and the process would reveal the matches that pass muster with both parties. Just as in a market, however, the system would lack the clarity of a well-written and coherent plan.

Another difference between centralized and decentralized systems resides in the structure of the networks that they form. Centralized systems tend to be characterized by tree-like hierarchies, while decentralized systems tend to self-organize into more distributed network topologies where cycles and shortcuts between nodes in different “branches” are also prevalent and where large hubs reduce the distance between any two nodes in the system.

The purpose of this paper is threefold. First, we describe some of the features and outcomes of the emergent structure of the aid network. Second, we study, in a very indirect and imperfect way, the degree to which the system has been able to arbitrage “supply” and “demand” for international development assistance. Third, we provide a relative ranking of the degree to which aid organizations, recipient countries and development issues are effectively coordinated. Finally, the paper also pretends to be a “proof of concept” regarding the potential use of the information available through the World Wide Web to analyze complex, non-market

activities more generally. Additionally, we have created the Aid Explorer, a web application providing information that we believe may help donors and recipients coordinate their efforts in a more decentralized and self-organizing manner.<sup>3</sup>

We do this by constructing a network that connects donor organizations, recipient countries and development issues. We analyze the degree to which this network allows countries to connect to organizations that share similar concerns, a property for which we develop a measure called Organizational Consistency. Finally, we use this to measure the degree to which organizations, countries and issues are adequately coordinated.

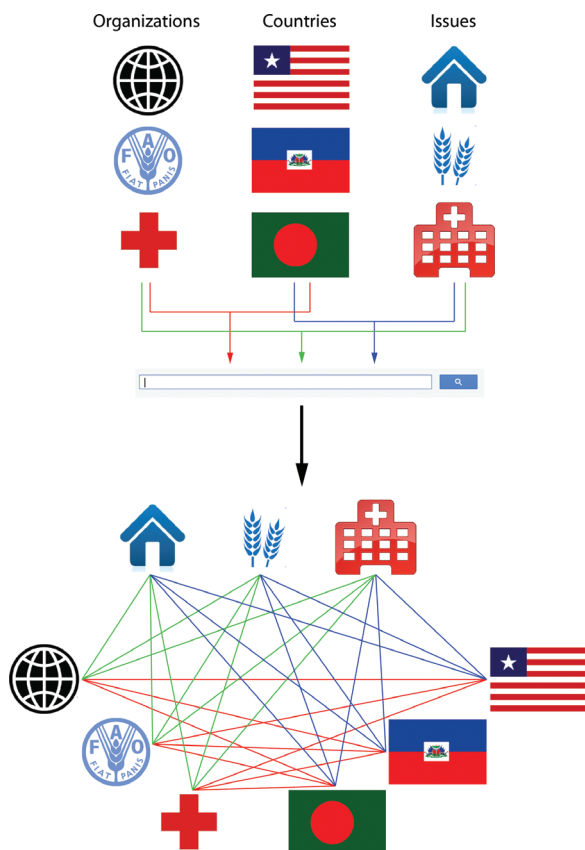
The study suffers from the limitations of its data source. We use the co-appearance of terms in the World Wide Web, which has been used to mine interesting associations in a number of applications, including gene and protein interactions (Cohen et al. 2005), social networks (Lee et al. 2010) and Mexican drug cartels (Coscia and Rios 2012). Words are notoriously flexible in their use, a fact that introduces noise to our data. They can also refer to negative rather than positive associations, as when an organization blames another for some bad outcome. In addition, our data does not have time resolution, meaning that we measure the connections in the data that is currently available on the web, independently of when the documents were written or published. Finally, we capture the intensity of speech, not the amount of money changing hands or the effectiveness of the projects undertaken. This implies that we cannot address issues regarding the duplication of efforts or the adequacy of the financial allocations. Also, the number of results for each query is approximate, because of optimizations made by search engines (Büttcher et al. 2010). Nevertheless, the results show that the method provides a good view of a system for which alternative empirical methodologies are not presently available.

### 3 Methodology

To analyze the state of aid coordination we study the tripartite network that connects aid organizations to each other and to recipient countries and development issues. Figure 1 depicts a representation of our methodology. We start from three distinct sets: the set of international aid organizations, the set of recipient countries and the set of development issues. For the set of organizations we include all the organizations that form part of the United Nations system, such as FAO or WHO; all the multilateral organizations including the Bretton Woods

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<sup>3</sup> <http://www.atlas.cid.harvard.edu/aidxp>



**Figure 1** A Representation of our workflow. Given a set of real world entities, we classify them into “Organizations,” “Countries” and “Issues.” We then make couples of elements from different classes and we feed the online search engine with these queries. The results are bipartite networks connecting each country with organization and issues, each organization with countries and issue and each issue with countries and organizations.

organizations (e.g., The World Bank, the IMF, the IFC, etc.) and the regional development banks; all the bilateral organizations (i.e., development aid organizations created by single countries, such as the Belgian Technical Cooperation Agency or the Italian Development Cooperation Program). We also include a fairly comprehensive list of private foundations, such as the Bill and Melinda Gates Foundations or the Soros Foundation, but exclude those that do not have an adequate web presence, defined by a threshold on the number of pages in their website (of more than 400 pages). The final list includes 153 organizations and is presented in Table 1.



**Table 1** The List of the Organizations Included in Our Work with their Corresponding Acronym when Present.

Institution	Acronym
Abu Dhabi Fund for Development	ADFD
Accion International	ACCION
Action Against Hunger	
Acumen Fund	
Adventist Development and Relief Agency	ADRA
African Capacity Building Foundation	ACBF-PACT
African Development Bank	AFDB
Aga Khan Development Network	AKDN
Agence d'Aide a la Cooperation Technique Et au Developpement	ACTED
Agencia Brasileira de Cooperacao	ABC
American Friends Service Committee	AFSC
American Jewish World Service	AJWS
American Red Cross	
American Refugee Committee	ARC
AmeriCares Disaster Relief and Humanitarian Aid Organization	AMERICARES
Amnesty International	
Andean Development Corporation	CAF
Arab Bank for Economic Development in Africa	BADEA
Arab Fund for Economic & Social Development	
Ashoka	
Asian Development Bank	ADB
Atlantic Philanthropies	
Australian Agency for International Development	AUSAID
Austrian Development Agency	AWSG
Austria Wirtschaftsservice Gesellschaft	
Belgian Policy Plan for Development Cooperation	
Belgian Technical Cooperation	
Bill and Melinda Gates Foundation	
Canadian International Development Agency	ACDI-CIDA
Caribbean Development Bank	CARIBANK
Carlos Slim Foundation	
Case Foundation	
Catholic Overseas Development Agency	
Childreach	
China Development Bank	CDB
China Development Industrial Bank	
Christian Aid	
Christian Reformed World Relief Committee	CRWRC
Church World Service	
Concern Worldwide	
Congo Basin Forest Fund	CBF-FUND
Cooperative for Assistance and Relief Everywhere	

(Table 1 Continued)

Institution	Acronym
Danish International Development Agency	
Deutsche Gesellschaft für Internationale Zusammenarbeit	GIZ
Development Alternatives Inc.	DAI
Direct Relief International	
Doctors Without Borders	
Dubai Cares	
Eurasia Foundation	
EuropeAid Development and Cooperation	
European Bank for Reconstruction and Development	EBRD
European Investment Bank	EIB
Fast Track Initiative Catalytic Fund	
Federal Ministry for Economic Cooperation and Development	BMZ
Finnish Department for International Development Co-operation	
Food and Agriculture Organization	FAO
Food For The Hungry	FH
Ford Foundation	
French Development Agency	
German Development Bank	KFW.DE
Global Alliance for Vaccines & Immunization	
Global Environment Facility	
Global Fund to Fight Aids, Tuberculosis and Malaria	
Google Org	
Grameen Bank	
Grameen Foundation	
Hellenic Aid	
Helvetas	
Hewlett Foundation	
High Commissioner of Human Rights	OHCHR
Human Rights Council	
Human Rights Watch	HRW
IBM International Foundation	
Instituto Portugues de Apoio ao Desenvolvimento	IPAD
Inter-American Development Bank	IADB
Inter-American Foundation	IAF
International Committee of the Red Cross	ICRC
International Cooperation and Development Fund	ICDF
International Development Research Centre	IDRC
International Fund for Agricultural Development	IFAD
International Monetary Fund	IMF
Irish Aid	
Islamic Development Bank	ISDB
Islamic Relief Worldwide	
Israel's Agency for International Development Cooperation	

(Table 1 Continued)

Institution	Acronym
Italian Development Cooperation Programme	
Japan Bank for International Cooperation	JBIC
Japan International Cooperation Agency	JICA
Japan Official Development Assistance	JODA
Joint United Nations Programme on HIV/AIDS	UNAIDS
Kauffman foundation	
Korea International Cooperation Agency	KOICA
Lemelson Foundation	
Liechtensteinische Entwicklungsdienst	
Life for Relief and Development	
Lutheran World Relief	LWR
Lux-Development	
MacArthur Foundation	
Maktoum Foundation	
Medair	
Medical Assistance Program International	MAP
Mercy Corps International	
Millennium Challenge Corporation	MCC
Mo Ibrahim Foundation	
Multilateral Investment Guarantee Agency	
Netherlands Ministry of Development Cooperation	
New Zealand Agency for International Development	
Nordic Development Fund	NDF
North American Development Bank	NADBANK
Norwegian Agency for Development Cooperation	NORAD
Novartis Foundation	
OPEC Fund for International Development	OFID
Oxfam International	
Poland Development Co-operation Department	
Refugees International	
Rockefeller Brothers Fund	RBF
Rockefeller Foundation	
Romania Official Development Assistance	AOD
Salvation Army International Headquarters	
Save the Children	
Schwab Foundation	
Seven fund	
Shell Foundation	
Skoll Foundation	
Slovak Aid	
Soros Foundation	
Spanish Agency for International Cooperation	AECID
Swedish International Development Cooperation Agency	SIDA

(Table 1 Continued)

Institution	Acronym
Swiss Agency for Development and Cooperation	SDC
Turkish International Cooperation and Development Agency	TIKA
UK Department for International Development	DFID
United Nations Capital Development Fund	UNCDF
United Nations Centre for Human Settlements	
United Nations Conference on Trade and Development	UNCTAD
United Nations Democracy Fund	
United Nations Development Fund for Women	
United Nations Development Programme	UNDP
United Nations Educational, Scientific and Cultural Organization	UNESCO
United Nations Environment Programme	UNEP
United Nations High Commissioner for Refugees	UNHCR
United Nations Office for Project Services	UNOPS
United Nations Relief and Works Agency for Palestine Refugees	UNRWA
United States Agency for International Development	USAID
United Way International	
US African Development Foundation	ADF
Waleed bin Talal Foundation	
West African Development Bank	BOAD
World Bank	WB
World Concern	
World Food Programme	WFP
World Health Organization	WHO
World Relief	
World Vision International	WVI

We construct the set of recipient countries by focusing on those with a population above 1 million. We then rank the list of remaining countries by GDP per capita at purchasing power parity and choose the poorest 110 countries. The richest country included is Argentina and the poorest country excluded is Russia. The final list of countries is included in Table 2.

Finally, the issues are defined through a manually curated list of keywords related to development. In this context, it is crucial to choose keywords that are used only in fairly specific contexts. For example, the term “Health” is ambiguous, because it may be used in reference both to medical and financial systems. Our list of issues is based on two types of keywords: concerns and activities. Concerns relate to goals or problems, such as poverty reduction, malnutrition or civil war; while activities relate to the kind of processes that

Table 2 The List of Countries Included in Our Work.

Afghanistan	Guinea	Niger
Albania	Guinea-Bissau	Nigeria
Algeria	Haiti	Pakistan
Angola	Honduras	Panama
Argentina	India	Papua New Guinea
Armenia	Indonesia	Paraguay
Azerbaijan	Iran	Peru
Bangladesh	Iraq	Philippines
Belarus	Ivory Coast	Republic of Congo
Benin	Jamaica	Romania
Bolivia	Jordan	Rwanda
Bosnia and Herzegovina	Kazakhstan	Senegal
Botswana	Kenya	Serbia
Brazil	Kosovo	Sierra Leone
Bulgaria	Kyrgyzstan	South Africa
Burkina Faso	Lao People Democratic	Sri Lanka
Burundi	Republic	Sudan
Cambodia	Latvia	Swaziland
Cameroon	Lebanon	Syria
Central African Republic	Lesotho	Tajikistan
Chad	Liberia	Tanzania
Chile	Libya	Thailand
China	Macedonia	Timor-Leste
Colombia	Madagascar	Togo
Costa Rica	Malawi	Tunisia
Democratic Republic of Congo	Malaysia	Turkey
Dominican Republic	Mali	Turkmenistan
Ecuador	Mauritania	Uganda
Egypt	Mauritius	Ukraine
El Salvador	Mexico	Uruguay
Eritrea	Moldova	Uzbekistan
Ethiopia	Mongolia	Venezuela
Gabon	Morocco	Vietnam
Gambia	Mozambique	Yemen
Georgia	Myanmar	Zambia
Ghana	Namibia	Zimbabwe
Guatemala	Nepal	
Guinea-Bissau	Nicaragua	

are carried out to achieve development goals, such as the banking system, the criminal justice or hospitals. Table 3 presents the list of issues, divided into concerns and activities.

**Table 3** The issues, divided into activities and concerns, included in our work.

Activities	Concerns
Agriculture	Poverty reduction
Banking System	Economic growth
Manufacturing	Rural development
Housing	Climate change
Transportation	Environmental sustainability
Electricity	Job creation
Sanitation	Technological development
Primary School	Homicides
Hospital	Civil war
Tourism	Natural disaster
Microenterprise	Gender
Small and Medium Enterprise	Democracy
Criminal Justice	HIV/AIDS
Reconstruction	Refugee
Humanitarian Assistance	School completion
	Infant mortality
	Malnutrition
	Human rights
	Corruption

We gather our data through two processes. First, we build a custom search engine that indexes only the websites of the 153 aid organizations, using the Google Custom Search API.<sup>4</sup> In this custom search engine, we are able to query either all pages or only those in the websites of the 153 aid organizations in search for documents that mention simultaneously the “World Bank” and “Medecins sans Frontiers,” or the “IMF” and “Poverty Reduction,” or the “African Development Bank” and Bangladesh. Acronyms and synonyms are handled by the entity recognition system of Google search engine. For the links that refer to organizations we searched the websites of the 153 aid organizations, but to study the relationship between a country and an issue we search the entire World Wide Web, and not only the websites of the aid organizations. We do this to better capture the importance of the issue for the country, beyond its salience in the aid community. As a measure of intensity of speech, we record the number of hits that contain each pair. This allows us to approximate the strength of the connection.

<sup>4</sup> <https://developers.google.com/custom-search/>

We search for pairs and not triplets or quadruplets, as this would increase the number of queries beyond what is feasible to do with the use of a public API.<sup>5</sup> For example, the set of quadruplets composed of pairs of organizations that cooperate on an issue in a country includes 86,977,440 possible combinations ( $153 \times 152 \times 110 \times 34$ ), which would constitute the number of Internet queries we would need to do to map out this space. At one query per second, which is the rate allowed by a non-special Google Custom Search Engine license, this translates to nearly 3 years of uninterrupted querying. We acknowledge this to be a limitation of our approach, since aid coordination usually includes more than two organizations, working in a set of countries and a set of issues, making relationships non-dyadic. These non-dyadic relationships, however, should still be expressed imperfectly in dyadic data, since triplets, quadruplets and other sub-graphs are composed of dyads.

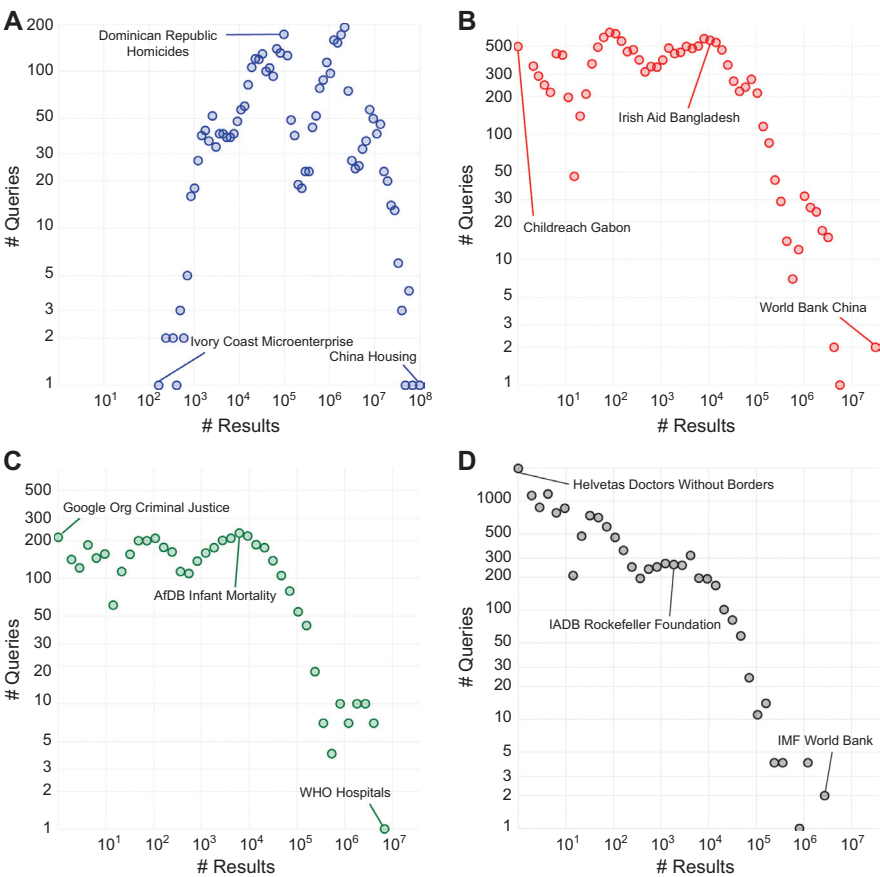
We begin by looking first at uni-partite networks, in which all nodes are of the same kind (connecting, for instance, countries to countries or organizations to organizations). In particular, we explore three such networks: the organization space, the issue space and the country space. In each of these, pairs of nodes are connected if both appear in the same documents with unexpectedly high frequency (defined below). Besides uni-partite networks, we also study the set of all possible bi-partite networks. A network is bi-partite when its nodes are of two different types and the links in the network connect only nodes of different type. In our case they are the Organization-Country (possible links are  $153 \times 110 = 16,830$ ), Organization-Issue ( $153 \times 34 = 5,202$ ) and Country-Issue ( $110 \times 34 = 3,740$ ) networks. These three bi-partite networks involve 25,772 observations. By looking at pairs, instead of triplets or quadruplets, we keep the data collection process manageable. In total, we performed 62,140 different queries.

Figure 2 reports the frequency distribution of hits for each of the queries performed. The figures show the number of queries that delivered a certain number of hits, or in network terms, the distribution of hits of each link. Because of the heterogeneous (fat-tailed) nature of these distributions, we present them in a double logarithmic scale and bin them using a technique known as log-binning. All of the empirically observed distributions are not well described by a simple analytical distribution, such as a power-law or a log-normal distribution. Nevertheless, they vary over several orders of magnitude, indicating that the distributions are broad, or fat-tailed.

To identify the most important links, we study their statistical significance. To do this we take contrast the frequency with which each link appears in our

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<sup>5</sup> A public web API is a tool that allows programs to access web content directly. These are commonly used to build dynamic content, but can also be used to retrieve data.



**Figure 2** The distributions of our results for the different classes of queries. (A) The distribution of results for the Country-Issue class of queries. (B) The distribution of results for the Organization-Country class of queries. (C) The distribution of results for the Organization-Issue class of queries. (D) The distribution of results for the Organization-Organization class of queries.

dataset and compare it with the number of times it would be expected to appear as a result of pure chance. We do this by taking the ratio between the observed and the randomly expected value, a measure called revealed comparative advantage  $R$  (Balassa 1965) in international trade, relative risk in statistics (Sistrom and Garvan 2004) or lift in computer science (Geng and Hamilton 2006). Consider  $N_{a,b}$  as the number of hits where organization  $a$  and country  $b$  appear together,  $N_a$  as the total number of hits in which organization  $a$  appears,  $N_b$  as the total number



of hits in which country  $b$  appears and  $N$  as the total number of hits in the Organization-Country queries. We define the *relevance* of a link between  $a$  and  $b$   $R_{a,b}$  as:

$$R_{a,b} = \frac{\frac{N_{a,b}}{N_a}}{\frac{N_{a,b}}{N_b}}.$$

$R_{a,b}=1$  implies that the number of hits obtained is exactly what would be expected by pure chance. Note that  $R_{a,b}$  controls for the number of total hits received by  $a$  and  $b$  and hence does not depend on the frequency with which different nodes appear in the data.  $R_{a,b}>1$  indicates that the link is unusually frequent and  $R<1$  that it is infrequent.

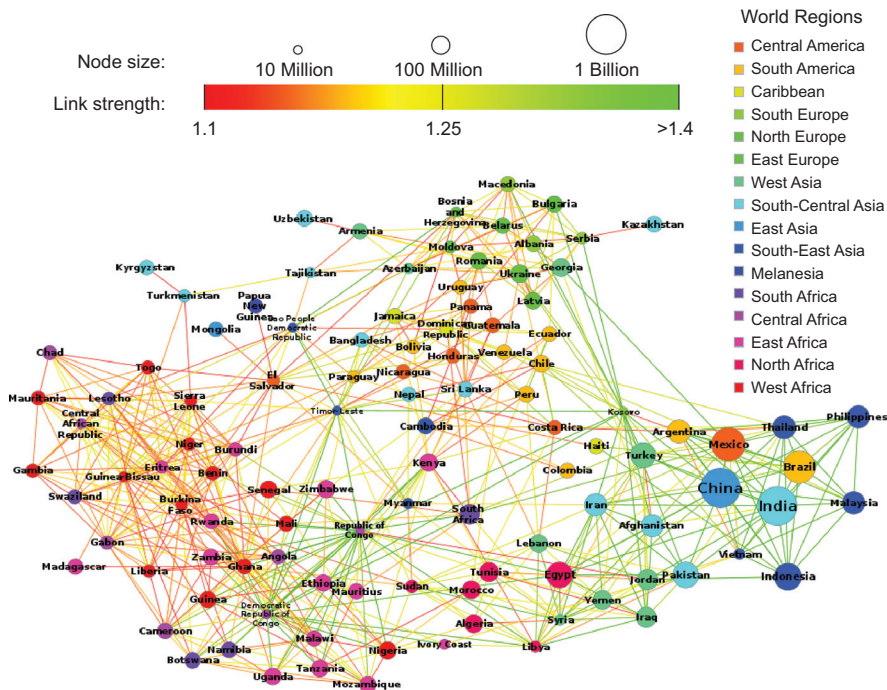
## 4 Results

### 4.1 Country Space

The Country Space is the network of country co-appearances in the websites of the 153 aid organizations. Figure 3 visualizes the network for  $R>1.1$  rather than 1.0, to capture the links that are more significant while also making the visualization less dense and more meaningful (only 10% of links are above 1.1). Here, the size of a node is proportional to the number of times the country is cited and its color reflects the region of the world the country belongs to. The color of the link is proportional to its relevance  $R$ , with green representing a stronger link and red a weaker link.

First, we observe that the space is connected, meaning that there are paths linking any two countries, despite the fact that we are only visualizing the strongest 10% of all links. This means that the aid community has not been split into unrelated clusters of countries but maintains some overall unity. Second, we observe a strong geographic effect: countries are much more likely to be connected with other countries in their same geographic region. This is particularly true in Eastern Europe, Central Asia, Sub-Saharan Africa, Middle East and North Africa and Latin America. Yet, the network also has a cluster of large countries, such as China, Brazil, India and Mexico, that concentrate a substantial number of citations and that belong to very different regions of the world.

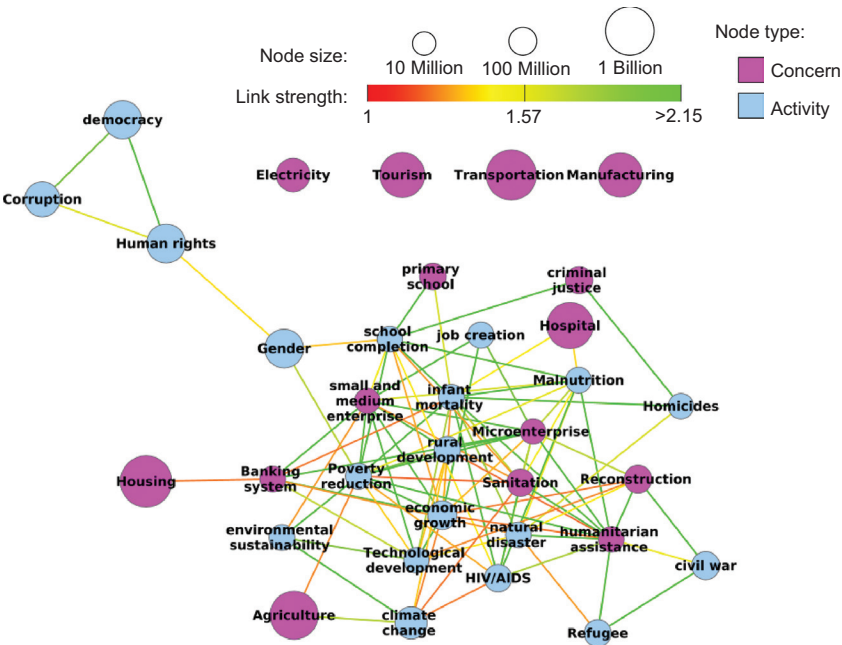
The structure of Figure 3 is suggestive of economies of scale. Large countries can be served by many organizations, independent of their relative location because their size may justify it. Smaller countries tend to be served through geographically based clusters.



**Figure 3** The Country Space, i.e., the uni-partite country network as a result of the co-appearances of each couple of countries, normalized using  $R$ . Each node is a country, connected if  $R$  was high. The color of the node is the world region where the country is located. The color of the edge is proportional to the  $R$  value (green=high, orange=medium, red=low). The size of the node is the amount of documents retrieved referring to the country.

## 4.2 Issue Space

Next, we show the issue space representing all 34 concerns and activities (Figure 4) connected according to the frequency with which organizations mention both in the same document. Here we find that when we include links with  $R_{i,o} > 1$  the space is not fully connected. Manufacturing, transportation, electricity and tourism are disconnected from the rest of the space, indicating that they do not tend to appear systematically together in documents discussing other issues related to economic and human development. This may reflect the fact that development thinking has tended to avoid industry-specific issues in growth strategies. Human rights, corruption and democracy form a cycle that



**Figure 4** The Issue Space, i.e., the uni-partite Issue networks as a result of the co-appearances of each couple of issues, normalized using the  $R$ . The nodes are the issues connected by their  $R$ . The color of the edge is proportional to the  $R$  value (green=high, orange=medium, red=low). Node size is proportional to the popularity of the issue in the set of results considered. Node color is used to distinguish concerns (purple) from activities (blue).

is weakly connected to the rest of the space. Refugees are linked to civil war, natural disasters and humanitarian assistance. Overall, the network illustrates a sensible picture of how the international community has organized itself around issues.

### 4.3 Organization Space

Figure 5 shows the Organization Space. Since in this case the number of possible links is too large for practical visualization, we start by linking nodes through a maximum spanning tree, using Kruskal's algorithm (Kruskal 1956). The algorithm starts by connecting the two most related organizations. Then it looks for the third organization that is more strongly related to either and connects it to the



number of possible links. We note that two of the 153 organizations, namely the Polish and the Romanian Development Agencies are not present in the network, as they are never cited together with other organizations. After constructing the maximum spanning tree, we add the strongest links so as to bring the average degree of the network to nearly 8. This provides 453 additional links and helps illustrate the communities that exist in this network. This is the same visualization technique used to map the Product Space (Hidalgo et al. 2007) and has also been used to visualize connections in the brain (Hagmann et al. 2008).

Here, we color organizations based on how specialized they are on certain issues. To do this we use Shannon's Entropy measure (Shannon 1948):

$$H(o) = -\sum_{i=1}^n p(o,i) \log_2(p(o,i)),$$

where  $o$  is an organization,  $i$  is an issue and  $p(o,i)$  is the relative frequency of issue  $i$  for organization  $o$ , or  $r_{o,i}/r_o$ .  $H$  should capture the fact that Human Rights Watch is more specialized than the World Bank. A high value of  $H$  indicates a high level of entropy, and therefore means that the organization deals with a more diverse set of issues. In fact  $2^H$  is equal to the effective number of issues that an organization is associated with (Jost 2006).

In Figure 5, nodes are sized according to their *betweenness centrality*. This is a network concept that measures the proportion of shortest paths between any two nodes that pass through a given node (Newman 2003, 2010). We interpret high betweenness centrality nodes -large nodes in our visualization – as important connectors, since these are organizations that lie in the shortest path linking many other organizations.

Figure 5 shows the effects of the broad distribution shown in Figure 2D. The space is held together by large hubs that show high betweenness centrality and high entropy. Networks with broad degree distributions can grow with very small increases in the average distance between the members of the network. This allows having an international community with a large number of organizations in which any two organizations are never too far apart. Interestingly the World Bank is the organization with the highest entropy and the most central of the nodes in the space. The second most central organization is Save the Children, which is also characterized by high entropy and centrality but connects preferentially to private donor organizations. The network exhibits a core-periphery structure where some large organizations, such as the World Bank, Save the Children, UNDP and The Ford Foundation, act as hubs and help keep the network together. We believe that it is no coincidence that the hubs have high issue entropy, since their lack of specialization is required to play a central role in the network.

We believe that this result is significant because a common discussion among development stakeholders is the degree to which organizations should specialize. Easterly (2007), for example, argues in favor of aid organizations specializing by country or issue in order to increase efficiency. Others have argued that a client-focused aid organization such as the World Bank, or the major regional development banks, should be able to work on the different issues that their heterogeneous developing-country partners face, and hence, must be able to deliver a more diverse suite of services. Our results suggest that, beyond this discussion, there is a systemic role played by the large, diverse organizations: they allow the network to remain connected so that the rising number of smaller, specialized entities can be linked to the rest by cooperating with the large, entropic organizations. Hence, the lack of specialization of the latter plays a systemic role. This implies that the degree of specialization of development agencies should be discussed taking into account the systemic role they play in the organization space.

#### 4.4 Bipartite Networks

We now move to discuss three bipartite networks. As mentioned earlier, a bipartite network has two kinds of nodes and the links in the network only connect nodes of different kinds. Our three bipartite networks are the Country-Issue, Organization-Issue and Country-Organization networks. Next, we compare the structure of each bipartite network to explore the ability of the system to achieve consistency in the coordination of which organizations operate in which countries and on which issues.

Tables 4, 5 and 6 show some of the most and least connected pairs for each of the three bipartite networks. It illustrates that the method does find reasonable connections. For example, for Organization-Issue queries we see a strong relation between “Accion International” and “Microenterprise” or “Dubai Cares” and “primary school,” which we expect, given the mandate of those institutions. At the same time, we see a weak connection between “Joint United Nations Programme on HIV/AIDS” and “Banking system,” or between “Doctors Without Borders” and “Tourism.” The same considerations can be done for the Country-Issue queries, where we find a strong association between “Mexico” and “criminal justice,” and “Malawi” and “HIV,” and weak connections between “Afghanistan” and “Small and Medium Enterprises.” Finally, we find a strong association between Country-Organization queries such as “Instituto Portugues de Apoio ao Desenvolvimento” and “Angola” and “North American Development Bank” and “Mexico,” and a weak link between the “Aga Khan Development Network” and “Uruguay.”

Table 4 Some examples of high and low  $R$  couplings of countries and issues.

Country	Issue	Results	$R$
Mexico	Criminal Justice	7,160,000	5.10753
Malawi	HIV/AIDS	820,000	4.74479
Ecuador	Microenterprise	8450	4.60329
...	...	...	...
Syria	School Completion	656	0.24633
Afghanistan	SME	2630	0.22421
Iran	Microenterprise	2320	0.18767

Table 5 Some examples of high and low  $R$  couplings of countries and organizations.

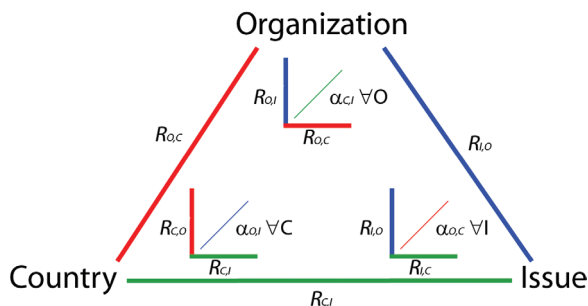
Organization	Country	Results	$R$
Instituto Portugues de Apoio ao Desenvolvimento	Angola	2530	62.74844
Arab Bank for Economic Development in Africa	Togo	8530	55.40399
North American Development Bank	Mexico	5320	22.91313
...	...	...	...
Aga Khan Development Network	Uruguay	39	0.01914
MacArthur Foundation	Bosnia and Herzegovina	141	0.01385
Google Org	Kazakhstan	5	0.00340

To study the tripartite network of organizations, countries and issues we propose to infer its properties by studying the relationship between the three possible bi-partite networks, as shown in Figure 6. Consider first the relationship between a country (e.g., Bolivia) and an issue (e.g., Microenterprise), which is captured as the bottom line of the triangle in Figure 6 and is measured by the

Table 6 Some examples of high and low  $R$  couplings of issues and organizations.

Organization	Issue	Results	$R$
Accion International	Microenterprise	2750	124.94252
Dubai Cares	Primary School	3980	46.32938
Kauffman Foundation	Job Creation	9920	16.35688
...	...	...	...
Amnesty International	Microenterprise	65	0.01047
Joint United Nations Programme on HIV/AIDS	Banking System	32	0.00410
Doctors Without Borders	Tourism	17	0.00067





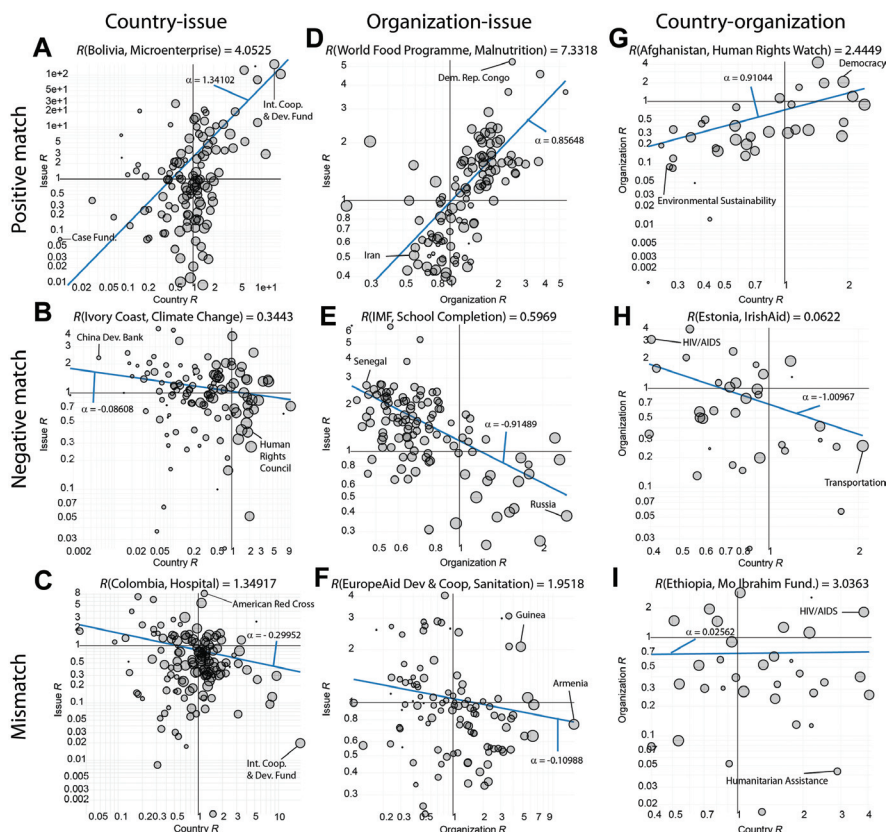
**Figure 6** A representation of our tripartite structure. We do not observe directly the tripartite structure, but each of the bipartite parts (Organization-Country, Organization-Issue and Country-Issue). To have an idea about how the tripartite structure organizes itself, for each direct bipartite observation (say the Country-Issue:  $R_{CI}$ ) we create an indirect observation via the other two edges, represented by a scattergram in the opposite vertex, that returns an  $\alpha_{CI}$  for the Country-Issue as the exponent of its power regression. This procedure is general for any tripartite structure observed only through its bipartite components.

associated relevance  $R_{CI}$  (equal to 4.05 in our example, as we shall see). This is the direct relationship measured by the number of documents that mention simultaneously Bolivia and Microenterprise in the World-Wide Web. Consider now the opposite vertex, meaning the top corner of the triangle. This vertex is about the relationship between the Organization-Issue and the Organization-Country networks. It shows how much the organizations that are interested in the country (e.g., Bolivia) are also interested in the issue (e.g., Microenterprise), as measured through the websites of the 153 development agencies. It is an indirect link between the country and the issue, created by looking at how the organizations relate to both.

This link is represented by a scattergram where the observations are organizations: the x-axis captures the degree to which each organization is related to Bolivia and the y-axis measures the degree to which each organization is related to microenterprise. The relationship is summarized by the slope of the scattergram, which we call *alignment* and label as  $\alpha_{CI}$ . In this example,  $\alpha_{CI}$  refers to the degree with which organizations that find Bolivia relevant are also interested in microenterprise.

Figure 7A–I shows the scattergrams suggested by Figure 6 for some selected cases. Figure 7A shows that organizations that are strongly (weakly) associated with Bolivia are also strongly (weakly) associated with the Microenterprise issue. For instance, the International Cooperation and Development Fund is strongly associated with both Bolivia and Microenterprise, whereas the Case Foundation is weakly associated with both Bolivia and Microenterprise. In this case we say





**Figure 7** Scatterplots of the three bi-partite networks, showing the relationship between the two complementary measures of relevance  $R$  for the Country-Issue, Organization-Issue and Organization-Country class of queries. The direct measure of relevance appears in the sub-title of the graph. The first row shows examples of combinations with high relevance  $R$ s and positive alignment  $\alpha$ ; in the second row shows examples of low  $R$ s and negative  $\alpha$ ; the third row shows examples of high but no significant alignment  $\alpha$ .

there is alignment because there is a positive association between the relevance of Bolivia and Microenterprise and this is captured by a high  $\alpha_{c,I}$ . The value of  $R_{c,I}$  is shown at the top of each graph in Figure 7.

Figure 7A–I shows 9 scattergrams arranged in three columns. The first column shows examples from the Country-Issue bi-partite network, the second column from the Organization-Issue network and the third from the Country-Organization network. Figure 7B shows an example of a negative association: in this case between Ivory Coast and climate change. The relevance of the direct relationship is low with an  $R_c = 0.34$ . We note that organizations associated with Ivory Coast are

not associated with climate change, and vice versa. By contrast, the bottom graph of this first column (Figure 7C) shows what looks like an incongruity between Colombia and hospital. Colombia is strongly connected to “hospital” ( $R_{c,I}=1.34$ ), but the organizations that are linked to Colombia are not those that focus on health. We call this a mismatch.

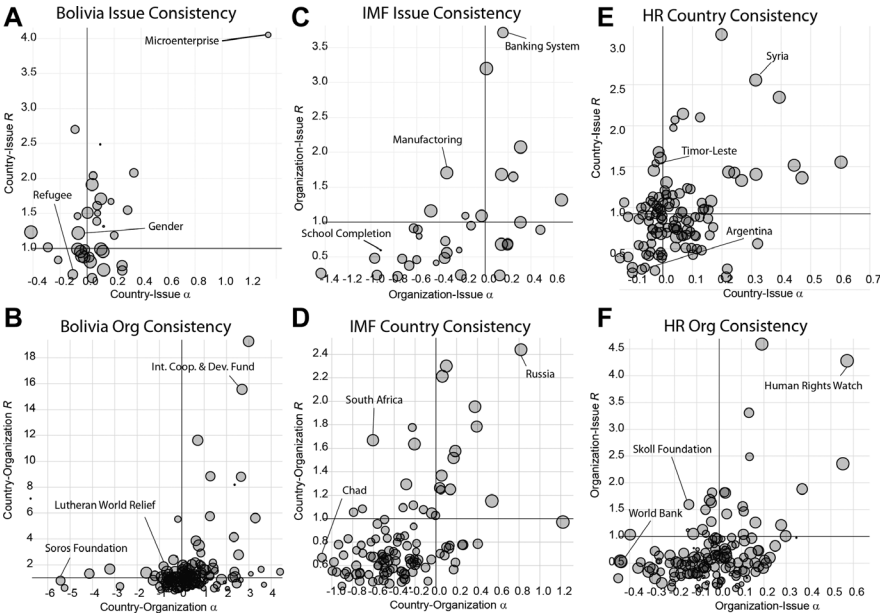
The middle column in Figure 7 (D–F) studies the links between organizations and issues, and the right column, the links between organizations and countries. In both cases we show one example with a positive association (Figure 7D and G), a negative association (Figure 7E and H) and a mismatch (Figure 7F and I). Given our tri-partite network, there are 25,772 possible scattergrams that can be constructed with our dataset and are available in the Aid Explorer, our web application.

## 4.5 An Intermediate View of the System

We can now look at a higher level of aggregation. We can look more systematically at the relationship between the relevance  $R_{i,j}$  and the alignment  $\alpha_{i,j}$  for  $i, j = (\text{Issue, Country, Organization})$  and  $j \neq i$  to assess a higher order degree of consistency of the aid coordination process. When  $R_{c,I}$  and  $\alpha_{c,I}$  are both high it means that the country is strongly related to the issue [as shown by the high  $R_{\downarrow}(C, I)$ ] but also that it is related preferentially to organizations that also focus significantly on the issue. We say that there is a positive match between relevance as captured by  $R_{c,I}$  and the alignment as captured by  $\alpha_{c,I}$ . We call this relationship the *issue-consistency* of the country.

Figure 8A represents the issue-consistency of Bolivia as it illustrates the relationship between relevance  $R_{c,I}$  and alignment  $\alpha_{c,I}$  for all issues and not just for microenterprise. Here, the y-axis is the relevance of the issue to Bolivia, measured by searching the whole World Wide Web. The x-axis measures the alignment  $\alpha_{c,I}$  between Bolivia and each issue through the consistency with which the organizations that are linked to Bolivia also find the issue salient, measured through the websites of aid organizations. For example, Figure 7A showed that the organizations associated with Bolivia are also associated with Microenterprise. Hence, in Figure 8A, microenterprise appears as a single point in the upper-right quadrant (high  $R_{c,I}$  and  $\alpha_{c,I}$ ). Conversely, Bolivia is not strongly associated with “Refugee” ( $R_{c,I} < 1$ ) and in this case the alignment of organizations associated with both Refugees and Bolivia is poor, with a negative  $\alpha_{c,I}$ . Overall, the Bolivia Issue Plot (Figure 8A) shows a high issue consistency of Bolivia, given that issues that are strongly associated with Bolivia (high  $R_{c,I}$ ) are also the issues where there is a positive alignment.

Figure 8A uses the relationship between the bottom of the triangle in Figure 6 and the top vertex. But we could also create a similar relationship using the left



**Figure 8** Consistency graphs: the scattergrams show relevance vs. alignment scatterplots for the three bi-partite plots and two consistency criteria. Given a country we plot the relationship between its issue relevance and the alignment  $\alpha$  of organizations with both the country and the issues. We repeat the same procedure also for organizations and issues using each of the two related concepts of alignment.

side of the triangle and the right-side vertex. This means that we would relate the relevance of the Organization-Country link  $R_{o,c}$  to the alignment between organization and country  $a_{o,c}$ , for a particular country, say Bolivia. This is shown in Figure 8B. Here the points on the scattergram are organizations rather than issues and it tells us which are the organizations that are salient to Bolivia and aligned with its issues. We refer to the relationship in Figure 8A as a country’s issue consistency, while that of Figure 8B as a country’s organizational consistency.

Figure 8C illustrates the issue consistency of organizations, using the IMF as an example. It is based on relating the right side with the left vertex of Figure 6. It shows which are the *issues* that are relevant to the IMF and that are aligned with it in terms of being important for the countries that the IMF cites more frequently. Figure 8D shows the country consistency of the IMF. It describes what are the countries that are relevant for the IMF and how aligned are the issues in those countries with those that the IMF focuses on. It is based on the left side and the right vertex of Figure 6.

Figure 8E looks at human rights as an example of the country consistency of issues. It is based on relating the bottom side to the top vertex of Figure 6. It

shows that Human Rights is a relevant issue in Syria and that the organizations that care about human rights also care about Syria, but it is a less salient issue in Argentina and so the organizations that care about human rights are less salient there. Figure 8F shows the organization consistency of issues. It is based on relating the right side to the left vertex of Figure 6. It shows that human rights are a salient issue for Human Rights Watch and that the organization focuses on countries where the human rights issue is salient. The opposite is true about the World Bank. By contrast, the Skoll Foundation finds the human rights issue relevant, but it does not align well with countries where the issue is salient and hence shows a negative alignment.

These sets of graphs can allow us to understand the aspects that make a country, an organization or an issue be more or less consistent. Given that there are three nodes and two types of consistency, we can generate  $2 \times 110 = 220$  country graphs,  $2 \times 153 = 306$  organization graphs and  $2 \times 34 = 68$  issue graphs, for a total of 594 graphs. This is a more aggregated view of the system than the 25,772 graphs of the kind shown in Figure 7, and can permit an analysis at a more systemic level, but it still is a very high dimensional description.

## 4.6 An Aggregate View of the Aid Coordination Network

To develop a more systemic view of the consistency of the aid coordination network we can re-aggregate the data to a higher level. Here, we look for a measure that tells us to what extent countries are being engaged by the right organizations, given the issues that are salient to them. Or alternatively, which organizations are engaged with the right countries, given the issues that they have decided to focus on. Or which are the issues that have an appropriate deployment of organizations in countries.

To make this judgment we propose to develop three overall consistency indexes, one for each node type in the tri-partite network. The index would be the proportion of the relevant links  $R_{ij} > 1$  that also have a consistency  $\alpha_{ij} > \xi$  where  $\xi$  is some cutoff value.<sup>6</sup> To make the indexes meaningful and comparable, we normalize them by the maximum number that the ratio could achieve on average for all

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<sup>6</sup> An alternative would be to consider a weighted measure, such as a correlation, rather than one based on cutoffs and counts. We did not take this route because we care about the quality of positive matches rather than the intensity of the negative matches. A regression approach treats both as symmetric. Secondly, correlation-based measures give higher weights to both positive and negative outliers and given the noise in the data, make the results less precise. This issue could be revisited if data based on triplets rather than pairs were used, as discussed in the concluding section.

the observations in the set if consistency was perfect. More precisely we propose to measure the following overall consistency indexes:

$$\kappa_{i,s} = \frac{\frac{N(\alpha_{i,j,s} > \xi \cap R_{i,j,s} > 1)}{N(R_{i,j,s} > 1)}}{\frac{N(\alpha_{i,j,s \forall j} > \xi)}{N(R_{i,j,s \forall j} > 1)}}$$

where the letter  $S$  refers to the set to which the elements belong,  $i$ =countries, organizations or issues. The index  $i$  refers to the specific country, issue or organization that is being considered and the  $j$ 's depend on the type of consistency that is considered. For the case of organizations, we use the country-consistency, as this will tell us, given the part of the world that the organization has chosen to focus on, whether it is in tune with the issues that are salient in those countries. We measure the degree to which the countries that the organization cares about, meaning that  $R_{o,c} > 1$ , are countries where the issues the organization cares about are also salient in the country, and hence have significant alignment  $\alpha_{o,c} > \xi$ .

For the case of countries we use their issue consistency, since we care that the issues that a country confronts be adequately tended to, independently of which organization does it. For issues we use the organization consistency, given that we want to assess the degree to which organizations are adequately deployed to deal with the issue that countries find salient.

In general, the number of total observations with  $\alpha_{i,j} > \xi$  will depend on the choice of  $\xi$ . To make our measure less dependent on this choice, we normalize the numerator dividing it by the ratio between the total number observations with  $\alpha_{i,j} > \xi$  and the number of observations with  $R_{i,j} > 1$ . This latter ratio is the highest possible value that perfect consistency would achieve, on average. This normalization factor makes a ratio of 1 for the whole class to imply perfect consistency.

To decide on the cutoffs, we note that out of 3740 observations in our Country-Issue network, 1768 have  $R_{c,i} > 1$  (47.3%) and 916 have a  $\alpha_{c,i} > 0.2$ . This means that if the system were perfectly coordinated we would get a ratio of 51.8% of the observations with  $R_{c,i} > 1$  that also have  $\alpha_{c,i} > 0.2$ . (see Table 7). So, by dividing by this ratio, we measure consistency relative to this maximum. This approach corrects for the arbitrariness of the cutoff and for noise in our data. We use the same cutoff of 0.2 for the organization-issue network and a cutoff of 0.5 for the Country-Organization network. The relevant numbers are shown in Table 7.

This measure allows us to get a sense of the degree to which each country, each organization and each issue is adequately coordinated. Tables 8, 9 and 10 show rankings based on our overall consistency measure.

**Table 7** Summary statistics on salience and consistency.

	(a)	(b)	(c)=(b)/(a)	(d)	(e)	(f)=(e)/(b)
Bipartite Networks	N	N (R>1)	Ratio	Cutoff	N (a>cutoff)	N (a)/N (R)
Country, Issue	3740	1768	47.3%	0.2	916	51.8%
Organization, Issue	5202	1889	36.3%	0.2	914	48.4%
Organization, Country	16,830	6890	40.9%	0.5	5206	75.6%

Table 8 shows the ranking of aid organizations according to our overall consistency index. Note that this measure considers the countries that the organization finds salient, so it should not penalize regionally focused institutions. It then asks whether the organization engages countries in issues that are salient to them. The top of the list is dominated by bilateral development agencies such as those of China, Norway, Portugal and Austria. However, this is not a general

**Table 8** The organization ranking. For each organization we report the degree to which the countries that are relevant for the organization also exhibit high issue alignment. The number reflects the percentage consistency relative to the theoretical maximum for the average of the whole table (see in the text and Table 7).

Rank	Organization	Country consistency
1	China Development Industrial Bank	132.35%
2	Childreach	132.35%
3	Norwegian Agency for Development Cooperation	127.25%
4	Andean Development Corporation	117.64%
5	Arab Bank for Economic Development in Africa	117.22%
6	Cooperative for Assistance and Relief Everywhere	117.00%
7	Fast Track Initiative Catalytic Fund	116.06%
8	Instituto Portugues de Apoio ao Desenvolvimento	115.80%
9	Austrian Development Agency	115.27%
10	International Cooperation and Development Fund	111.45%
11	Austria Wirtschaftsservice Gesellschaft	109.24%
12	West African Development Bank	108.99%
13	New Zealand Agency for International Development	108.86%
14	German Development Bank	108.86%
15	Korea International Cooperation Agency	108.28%
16	African Capacity Building Foundation	107.53%
17	United Nations Capital Development Fund	106.22%
18	UK Department for International Development	105.88%
19	Belgian Technical Cooperation	105.20%
20	EuropeAid Development and Cooperation	103.99%

(Table 8 Continued)

Rank	Organization	Country consistency
21	OPEC Fund for International Development	103.80%
22	Nordic Development Fund	103.73%
23	US African Development Foundation	102.94%
24	Spanish Agency for International Cooperation	102.74%
25	Japan Bank for International Cooperation	101.80%
26	Adventist development and relief agency	100.26%
27	Christian Reformed World Relief Committee	99.77%
28	United States Agency for International Development	99.67%
29	Swedish International Development Cooperation Agency	99.26%
30	Congo Basin Forest Fund	98.70%
31	Japan Official Development Assistance	95.58%
32	Mercy Corps International	95.06%
33	Accion International	93.92%
34	United Way International	93.66%
35	Canadian International Development Agency	91.62%
36	Global Alliance for Vaccines & Immunization	90.99%
37	Agencia Brasileira de Cooperacao	90.55%
38	Development Alternatives Inc.	90.38%
39	International Fund for Agricultural Development	89.29%
40	Medical Assistance Program International	88.23%
41	Italian Development Cooperation Programme	86.63%
42	Hellenic Aid	84.90%
43	Grameen Foundation	84.84%
44	Israel's Agency for International Development Cooperation	84.22%
45	American Refugee Committee	83.33%
46	Asian Development Bank	80.88%
47	United Nations Development Programme	79.41%
48	Case Foundation	79.41%
49	Inter-American Foundation	78.20%
50	Acumen Fund	76.85%
51	Islamic Relief Worldwide	76.62%
52	Lux-Development	76.10%
53	Food For The Hungry	75.00%
54	United Nations Relief and Works Agency for Palestine Refugees	74.24%
55	French Development Agency	73.53%
56	AmeriCares Disaster Relief and Humanitarian Aid Organization	73.53%
57	Swiss Agency for Development and Cooperation	72.70%
58	Slovak Aid	71.84%
59	Aga Khan Development Network	70.58%
60	Global Fund to Fight Aids, Tuberculosis and Malaria	68.01%
61	Japan International Cooperation Agency	67.58%
62	World Concern	66.17%
63	Novartis Foundation	66.17%
64	World Vision International	65.29%



(Table 8 Continued)

Rank	Organization	Country consistency
65	Global Environment Facility	64.63%
66	Life for Relief and Development	63.89%
67	American Red Cross	63.89%
68	Danish International Development Agency	62.04%
69	Multilateral Investment Guarantee Agency	61.64%
70	Church World Service	61.33%
71	United Nations Centre for Human Settlements	61.23%
72	Irish Aid	60.56%
73	World Food Programme	59.39%
74	United Nations Development Fund for Women	58.82%
75	Medair	58.82%
76	American Jewish World Service	58.39%
77	Australian Agency for International Development	55.50%
78	Shell Foundation	54.50%
79	Schwab Foundation	54.50%
80	Lemelson Foundation	53.92%
81	Caribbean Development Bank	53.77%
82	Atlantic Philanthropies	52.94%
83	North American Development Bank	50.90%
84	Netherlands Ministry of Development Cooperation	50.60%
85	European Investment Bank	49.94%
86	Lutheran World Relief	49.16%
87	Joint United Nations Programme on HIV/AIDS	48.76%
88	African Development Bank	48.53%
89	IBM International Foundation	47.27%
90	Dubai Cares	46.32%
91	Rockefeller Brothers Fund	46.03%
92	Carlos Slim Foundation	44.12%
93	Refugees International	43.53%
94	Helvetas	41.79%
95	United Nations High Commissioner for Refugees	41.72%
96	World Relief	41.66%
97	Arab Fund for Economic & Social Development	39.70%
98	Amnesty International	39.70%
99	International Committee of the Red Cross	39.21%
100	Food and Agriculture Organization	39.16%
101	Catholic Overseas Development Agency	38.50%
102	High Commissioner of Human Rights	38.47%
103	Seven fund	37.81%
104	Direct Relief International	37.81%
105	Oxfam International	37.53%
106	American Friends Service Committee	37.22%
107	China Development Bank	36.76%
108	United Nations Environment Programme	36.61%



(Table 8 Continued)

Rank	Organization	Country consistency
109	International Development Research Centre	34.83%
110	Eurasia Foundation	34.83%
111	Human Rights Watch	34.66%
112	Millennium Challenge Corporation	34.61%
113	United Nations Office for Project Services	34.41%
114	Bill and Melinda Gates Foundation	33.61%
115	Deutsche Gesellschaft für Internationale Zusammenarbeit	32.28%
116	Concern Worldwide	30.08%
117	Inter-American Development Bank	29.88%
118	Skoll Foundation	29.41%
119	Mo Ibrahim Foundation	29.05%
120	Human Rights Council	28.95%
121	Grameen Bank	28.77%
122	Waleed bin Talal Foundation	22.06%
123	Turkish International Cooperation and Development Agency	21.61%
124	Action Against Hunger	21.61%
125	United Nations Conference on Trade and Development	21.10%
126	Soros Foundation	20.05%
127	European Bank for Reconstruction and Development	20.05%
128	Agence d'Aide à la Coopération Technique Et au Développement	17.08%
129	MacArthur Foundation	16.54%
130	Google Org	16.54%
131	Federal Ministry for Economic Cooperation and Development	16.54%
132	Doctors Without Borders	16.29%
133	Abu Dhabi Fund for Development	13.57%
134	World Health Organization	13.23%
135	United Nations Democracy Fund	12.22%
136	United Nations Educational, Scientific and Cultural Organization	12.17%
137	International Monetary Fund	11.51%
138	Finnish Department for International Development Co-operation	11.03%
139	Christian Aid	10.18%
140	Save the Children	9.57%
141	Salvation Army International Headquarters	9.45%
142	Maktoum Foundation	8.82%
143	Ashoka	5.29%
144	Liechtensteinische Entwicklungsdienst	5.09%
145	Kauffman foundation	4.01%
146	Hewlett Foundation	2.94%
147	Rockefeller Foundation	2.45%
148	Ford Foundation	1.59%
149	World Bank	0.00%
150	Islamic Development Bank	0.00%
151	Belgian Policy Plan for Development Cooperation	0.00%

**Table 9** The country ranking. For each country we report the degree to which the issues that are relevant for the country also exhibit high organizational alignment. The number reflects the percentage consistency relative to the theoretical maximum for the average of the whole table (see in the text and Table 7).

Rank	Country	Issue consistency
1	El Salvador	124.89%
2	Iraq	122.83%
3	Mauritius	120.63%
4	Chile	115.81%
5	Iran	107.23%
6	Liberia	105.28%
7	Syria	96.51%
8	Panama	96.51%
9	Pakistan	96.51%
10	Afghanistan	96.51%
11	Gambia	91.43%
12	Benin	91.43%
13	Bosnia and Herzegovina	90.07%
14	Serbia	87.73%
15	Albania	87.73%
16	Cameroon	86.86%
17	Tunisia	85.78%
18	Macedonia	85.78%
19	Jamaica	85.78%
20	Ecuador	85.78%
21	Mauritania	84.44%
22	Sri Lanka	82.72%
23	Niger	82.72%
24	Sierra Leone	81.27%
25	Mexico	80.42%
26	Argentina	80.42%
27	Papua New Guinea	79.48%
28	Myanmar	78.96%
29	Lebanon	77.21%
30	Jordan	77.21%
31	Honduras	77.21%
32	Botswana	77.21%
33	Ghana	75.06%
34	Guinea-Bissau	74.24%
35	Costa Rica	74.24%
36	Zambia	73.53%
37	Nicaragua	73.53%
38	Yemen	72.38%
39	Romania	72.38%
40	Uruguay	71.11%

(Table 9 Continued)

Rank	Country	Issue consistency
41	Paraguay	71.11%
42	Armenia	71.11%
43	Dominican Republic	70.19%
44	Ukraine	68.93%
45	Ivory Coast	68.93%
46	Kazakhstan	68.12%
47	Gabon	68.12%
48	Chad	68.12%
49	Azerbaijan	68.12%
50	Republic of Congo	67.13%
51	Peru	64.34%
52	Namibia	64.34%
53	Mozambique	64.34%
54	Morocco	64.34%
55	Mali	64.34%
56	Lao People Democratic Republic	64.34%
57	Haiti	64.34%
58	Colombia	64.34%
59	Burkina Faso	64.34%
60	Algeria	64.34%
61	Senegal	61.41%
62	Togo	60.95%
63	Tajikistan	60.32%
64	Malawi	58.74%
65	Latvia	57.90%
66	Cambodia	57.90%
67	Bulgaria	57.90%
68	Madagascar	56.77%
69	Central African Republic	56.77%
70	Guinea	56.30%
71	Libya	55.15%
72	Ethiopia	55.15%
73	China	55.15%
74	Brazil	55.15%
75	Democratic Republic of Congo	54.04%
76	Guatemala	53.61%
77	Nepal	50.79%
78	Zimbabwe	48.25%
79	Swaziland	48.25%
80	Philippines	48.25%
81	Eritrea	48.25%
82	Timor-Leste	46.32%
83	Nigeria	45.41%
84	India	45.41%

(Table 9 Continued)

Rank	Country	Issue consistency
85	Angola	45.41%
86	Moldova	44.54%
87	Bangladesh	43.87%
88	Venezuela	42.89%
89	Uganda	42.89%
90	Sudan	42.89%
91	Lesotho	40.63%
92	Bolivia	40.63%
93	Kosovo	35.09%
94	Burundi	35.09%
95	Turkmenistan	29.69%
96	South Africa	28.95%
97	Rwanda	28.95%
98	Malaysia	27.57%
99	Kenya	27.57%
100	Egypt	27.57%
101	Uzbekistan	24.13%
102	Kyrgyzstan	24.13%
103	Belarus	24.13%
104	Tanzania	19.30%
105	Georgia	17.55%
106	Indonesia	14.85%
107	Vietnam	12.87%
108	Turkey	0.00%
109	Thailand	0.00%
110	Mongolia	0.00%

feature of bilateral agencies as those of Belgium, Finland and Abu Dhabi are at the bottom of the list. At the top of the list we do find some multilateral development agencies such as the Arab Bank for Economic Development in Africa and the Andean Development Corporation, although the Islamic Development Bank is at the bottom of the list. But the large central hubs such as the World Bank, the IMF and Save the Children are at the bottom of the list. This may be due to their high entropy, as they deal with many countries and issues, given the systemic role we have argued they perform.

We find many private foundations at the bottom of the list, such as the Ford, Rockefeller, Hewlett, Kauffman, and Google Foundations. This may be due to the fact that these organizations are not necessarily focused on typical development issues. Some as in the case of the Soros Foundation, tend to take a contrarian

**Table 10** The issue ranking. For each organization we report the degree to which the issues that are relevant for countries also exhibit high organizational alignment. The number reflects the percentage consistency relative to the theoretical maximum for the average of the whole table (see in the text and Table 7).

Rank	Issue	Organization Consistency
1	Malnutrition	129.16%
2	School completion	127.75%
3	Poverty reduction	127.39%
4	Humanitarian assistance	125.47%
5	HIV/AIDS	125.01%
6	Environmental sustainability	118.09%
7	Infant mortality	114.40%
8	Microenterprise	112.39%
9	Homicides	103.33%
10	Small and medium enterprise	101.69%
11	Refugee	99.50%
12	Sanitation	99.19%
13	Natural disaster	93.58%
14	Rural development	85.51%
15	Criminal justice	79.01%
16	Primary school	55.76%
17	Technological development	46.50%
18	Gender	40.82%
19	Human rights	39.74%
20	Civil war	39.74%
21	Electricity	36.74%
22	Agriculture	36.04%
23	Transportation	35.18%
24	Democracy	24.31%
25	Economic growth	16.31%
26	Hospital	15.12%
27	Manufacturing	13.78%
28	Climate change	11.81%
29	Tourism	8.98%
30	Corruption	8.61%
31	Reconstruction	7.75%
32	Housing	7.65%
33	Job creation	3.50%
34	Banking system	0.00%

view. Nevertheless, Childreach is at the top of the list. According to our rankings, the average level of consistency is 59.2% of the theoretical maximum described above.

Table 9 shows the ranking of countries. Here we used the issue consistency meaning that we look at the issues that are salient in the country and ask whether the organizations that are involved with the country also find those issues salient. There is a very broad distribution of values between countries of different sizes, regions and levels of development at both ends of the list. In this case, the average level of consistency is 62.1% of the theoretical maximum.

Table 10 shows the ranking of issues. Here we use the organizational consistency, meaning that we look at the organizations that find the issues salient and ask whether they are involved in the countries that are also strongly linked to those issues. Here the issues that have to do with poverty such as malnutrition, poverty reduction, humanitarian assistance and infant mortality appear high on the list, while the issues that deal with economic growth, such as banking systems, job creation or manufacturing, are at the bottom. Interestingly, environmental sustainability appears to be better coordinated than global warming, while human rights, technological development and gender appear in the middle of the list. For the issues, we observe an average consistency equal to 61.4% of the theoretical maximum.

## 5 Concluding Remarks

International development is a highly complex global enterprise that must confront coordination problems of paramount proportions. Since World War II, the proliferation of states, organizations and issues has created a space that is large, complex and rapidly evolving. In this paper we have developed a method that leverages the vast amount of information available on the web to measure the tripartite network connecting countries, organizations, and issues, and use it to estimate the degree of coordination that the world has, *de facto*, been able to achieve.

Given a set of countries and an evolving set of issues, coordination requires the ability of organizations to adapt either the portfolio of countries that they serve, or the issues they work on. Our measures indicate that the international aid network has indeed achieved a significant level of *de facto* coordination, as expressed in their online speech. Our methods, however, show that in many cases coordination is low, highlighting countries that are poorly served, issues that are not well attended, and organizations that appear to be focusing on the wrong combination of places and issues. Donors and recipients can use the tools and information we developed in this paper and in the Aid Explorer, the associated web application, to identify, in a decentralized manner, areas for improvement.

During the 1980s and 1990s, aid coordination was led by the World Bank. Since then, the *de jure* approach to aid coordination has changed and it is now based on the Millennium Development Goals (2000), the Paris Declaration for Aid Effectiveness (2005) and the Accra Agenda for Action (2008). It implies a major effort at decentralizing decision-making by giving a more prominent role to recipient national governments. The idea is that these governments should develop their own poverty reduction strategy in accordance with the MDGs and these strategies should be the basis for a formal coordination process organized through consultative groups and roundtables.

This approach has the advantage of increasing the number of coordinators (from say one – i.e., The World Bank – to the number of recipient governments), thus reducing the amount of coordination that each must achieve, but it still leaves quite a complex task. After all, there are hundreds of aid organizations, hundreds of recipient state and local governments, dozens of executing agencies and dozens of issues that need to be paired.

Under these conditions, self-organizing processes may exhibit some distinct advantages. First, they require little *ex ante* explicit coordination. Second, they have the power to reveal information about emerging needs and organizational expertise. Third, they can deal with levels of complexity that strict hierarchies have trouble coping with. Yet, self-organizing processes also have limitations, since they need to satisfy some basic requirements for them to be able to operate properly.

First and foremost is information: donors must be able to know what recipients want and what other donors are willing to supply. Recipients must be able to know what donors are willing to provide and what they might be willing to consider. If donors had this information, and if they were focused on maximizing the bang of their development bucks, given whatever goals or preferences they have, they will naturally move towards the needs that are least attended among those that they care about. By the same token, if a recipient country knew about the willingness of each donor to support different activities they would adjust their priorities and partnerships to achieve maximum effect, given their preferences. So information is key for the self-organizing process to be efficient.

Second, a system based on matches between willing donors and willing recipients is bound to increase the number of transactions, and if these have very large transaction costs, then it might be better to forgo this option in favor of a more centralized process that avoids the multiplicity of transactions. The international development community has *de facto* understood this problem and the Accra Agenda for Action calls for streamlining procurement rules and reporting standards in order to reduce the recipient's transactions costs associated with working with multiple partners.

In this context, a system that highlights the countries that are being poorly served, the issues that are inadequately attended and the organizations that have an imbalance between their goals and their partners may allow the different players of the aid community to improve their effectiveness without the need for a central planner. Such a system may allow supply and demand to find each other more quickly and for organizations and countries to change strategies more dynamically in the face of new information. This paper presents a very crude first approach to collect the data, and develop the analytics, that would inform such a system.

Our approach does approximate what is required for decentralized coordination because it facilitates the identification of the issues and countries that organizations care about and the issues that countries care about. Our metrics can be used as “prices” that facilitate unrealized matches between willing donors and recipients: Organizations and countries will seek to interact with partners that share interests, in the same way as in the market the equilibrium involves matching willing sellers with willing buyers. The coordination occurs *ex post* through the matching process rather than *ex ante* through some assignment of roles and tasks, something that is very hard to achieve given the large number of possible combinations and the fact that donor agencies are not hierarchically related and so need not follow each other’s orders. Our associated website, The Aid Explorer, should be useful in identifying potential new matches and hence accelerate the decentralized coordination process.

While our approach is an approximation, it could be much improved. For example, we have queried only pairs and not triplets or quadruplets. We were deterred from this more exhaustive approach by the limitations imposed by public APIs and by out of pocket search costs of such an endeavor, given the constraints of academic research. But a system that is supposed to guide hundreds of billions of dollars of support should find the additional costs of better data quite insignificant. Moreover, if the data gathering process was done continuously, time resolution could be improved. In addition the issues that the analysis considers may emerge from “word cloud” analysis of the organizations’ websites, rather than from a manually curated list. Finally, the number of organizations, both on the donor and the recipient sides, could be greatly expanded. While we believe that our simplified approach has shown that it has value as a proof-of-concept, a full implementation of the above agenda would transform it into a comprehensive tool for professional and institutional use.

One word of caution that needs to be considered is the possibility that organizations will try to game the rankings provided by a system that is based mostly on the collection of online data. In a world where the expression of an organization in the web is considered an important indicator of their development expertise,



there are clear incentives for an organization to cheat by focusing on “making noise” about development assistance instead of actually providing it. Ultimately, organizations looking to get ahead of others might try to inundate the web with documents that mention the keywords used in the data collection process. While this is an important caveat for the future applicability of these methods, it is important to note that this caveat does not apply exclusively to the methods presented in this paper. In fact, the incentive to focus on providing signals, instead of performing actions, will always be present in a world where it is hard to attribute outcomes to individual actors.

The methodology we developed in this paper can be applied to other contexts. The most similar situation is that of private foundations and their recipients. Much more ambitious would be the analysis of the interactions between government agencies and between these and the other social organizations such as corporations, unions and other non-governmental entities. Here, governments manage  $\sim 10^6$  pages of legislations and  $\sim 10^3$  executing agencies. They face hundreds of thousands of organizations that are affected by this complex hyperspace. Under these conditions, there must be many Pareto improvements that are impeded by the information problem that this structure faces. Tools that relax this information problem by exploiting the web may allow for a more efficient decentralized coordination in complex social systems.

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