Measuring Transparency*

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Abstract

Transparency is often viewed as crucial to government accountability, but its measurement remains elusive. This concept encompasses many dimensions, which have distinct effects. In this paper, we focus on a specific dimension of transparency: governments’ collection and dissemination of aggregate data. We construct a measure of government transparency, using an item response model that treats transparency as a latent predictor of the reporting of data to the World Bank’s World Development Indicators. The resultant index covers 149 countries from 1980-2008. Our measure, the

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HRV Index (1) systematically rates low to middle income autocracies as more transparent than does Freedom House, and (2) rates democracies – particularly high-income democracies – as more transparent than does the measure based on newspaper circulation. We demonstrate that HRV is a stronger predictor of a wide range of governance outcomes than is media circulation, particularly for autocratic regimes.
Transparency, broadly defined, relates to the full flow of information within a polity. A lengthy literature stresses the importance of the provision of information to the populace to ensure the accountability of government, and thus emphasizes the role of transparency as a determinant of government performance (see, for instance Adserà, Boix and Payne, 2003; Besley and Burgess, 2002).

This broad notion of transparency, however, may encompass many different forms of information transmission. The transmission of information to the citizenry is affected by the freedom, structure and size of the mass media market (Adserà, Boix and Payne, 2003; Besley and Burgess, 2002; Djankov et al., 2003). Transparency may also reflect the willingness of government bodies to provide information regarding their decision-making processes (Stasavage, 2003). Institutional structures, such as the presence or absence of Freedom of Information Laws (FOILs) may ease or hinder the flow of information to the public (Islam, 2006). One may even conceive of social capital and/or ethnic divisions as affecting the flow of information between citizens, thus aiding or hindering their ability to hold political agents to account (Habyarimana et al., 2009; Grief, 2006).

This multifaceted nature of transparency poses difficulties for measurement, theory, and theory testing. Optimally, theoretical models should specify which forms of information transmission are most relevant to the mechanism being studied. Empirical measures should be devised to precisely capture specific dimensions of transparency. And empirical tests of theories should employ these precise measures – and control for alternative information transmission mechanisms. By contrast, most existing findings rely on the broad claim that transparency is relevant to accountability, and support this claim using proxies for information transmission. Such statements fail to precisely identify mechanisms and may misstate the effect of specific informational channels.

In this paper, we take one step towards clarifying theoretical mechanisms and empirical measures. We introduce an index of a particular dimension of transparency: the collection
and dissemination of credible economic data by national statistical offices. To construct this measure, we rely on a Bayesian item response model which treats transparency (along this dimension) as a latent predictor of the reporting or non-reporting of data in the World Bank’s World Development Indicators (WDI) data series. Our model analyzes 172 items corresponding to the 172 variables relating to Economic Policy and Debt consistently collected by the WDI over time.\footnote{We defer to the WDI’s classification of variables. Other categories include Education, Environment, Financial Sector, Health, Infrastructure, Labor and Social Protection, Poverty, Private Sector and Trade, and Public Sector. See \url{http://databank.worldbank.org/ddp/home.do}. We refer to variables as consistently collected over time if they are reported by at least one country in every year contained in the WDI.} Since the WDI obtains its data from other international agencies that, in turn, obtain their data from national statistical offices, our measure is a valid indicator of governments’ efforts to collect and disseminate economically relevant information. Moreover, because the WDI omits data considered ‘questionable’, our index reflects the collection and dissemination of credible information.\footnote{See the World Bank statements about World Development Indicators: \url{http://data.worldbank.org/about/data-programs}, accessed March 7, 2011. In some instances, these data are omitted by the WDI. In others, the international agencies that provide information to the WDI weed out questionable data earlier in the collection process.}

Our index thus focuses on a specific dimension of transparency largely ignored in the literature.\footnote{For a notable exception, see Islam (2006) who focuses on the speed with which data are reported. Other exceptions include Bueno de Mesquita et al. (2003), Lebovic (2006), and Stone (2008).} This measure is based on objective criteria, rather than on subjective expert judgments. Moreover, our reliance on the presence or absence of data from a long-running data series provides greater coverage than many commonly used alternatives. We construct transparency measures for 149 countries covering a period from 1980-2008. Our index is also measured at the interval level – by construction, a unit change in our scale has a consistent meaning regardless of value of the measurement.

In addition to constructing an index of transparency, we offer a theoretical argument as to why this dimension of transparency is important to government accountability. Information collected and disseminated by the government has a key property: that of aggregation.

\footnote{\url{http://databank.worldbank.org/ddp/home.do}}
Because of the high fixed costs involved in collecting aggregate data, and because of the public goods-like property of such information (Rodrik, 1995), the government is the only entity able to assemble data on such matters as economic well-being, price levels, growth, etc. This property of aggregation is crucial for two reasons: (1) By virtue of the law of large numbers, broad aggregates are likely to more accurately reflect the efficacy of government policies than are personal experiences or news reports. When citizens are uncertain as to the appropriateness of different policy choices, the presence or absence of aggregate data is likely to prove crucial in ensuring government accountability. (2) Information about aggregate outcomes may also facilitate collective action. Poor aggregate outcomes are likely to indicate that dissatisfaction with the government is widespread. This knowledge may facilitate collective activities – such as protest or revolt – to discipline the government for its poor behavior in a way that an active press or widespread social contacts is unlikely to achieve.

We thus offer a precise measure of a limited but important component of transparency and suggest its theoretical relevance to government accountability. We emphasize that our measure does not capture all the dimensions of transparency, that accountability itself is complex and influenced by many factors. Moreover, data dissemination is related to other factors beyond government accountability, notably a country's level of economic development. Still, our measure is crisply defined, reproducible, meaningful, and important as both an intrinsically interesting dependent variable and as a potential explanatory variable. We encourage scholars to use it – along with other appropriate measures of other facets of transparency – to test various hypotheses about the relationships between transparency and accountability, governance, and democracy. Will will make our index publicly available on publication of this research.
1 Existing Literature

A large literature in political science relates transparency to government accountability. These theories predominantly build upon retrospective voting models (Barro, 1973; Ferjohn, 1986), in which citizens discipline the government by the threat of removal from office. Information is modeled as improving the precision (reducing the variance) with which the public observes either the outcome of government decisions or the government’s policy choice (Besley, 2006). Such studies typically conclude that improvements in the flow of information enable citizens to adopt superior strategies for disciplining politicians, thereby improving government performance. Empirical studies (e.g. Adserà, Boix and Payne, 2003; Besley and Burgess, 2002) have lent support to these conclusions.

The precise type of information being transmitted may, however, be crucial. In the models discussed above, the welfare implications of transparency depend on whether the citizenry better observes the outcome of a government policy or if the public better observes the policy choice. For instance, Prat (2005) demonstrates that information regarding the policy choice of the government can be detrimental when the public is imperfectly informed of the optimal policy and when governments are affected by career concerns. Career concerns may cause governments to adopt conventionally accepted policies, regardless of any private information, to avoid being seen as incompetent – and suffering the resultant risk of replacement. Similarly, Stasavage (2004) demonstrates that transparency regarding the conduct of international negotiations may lead to increased posturing by government representatives, even if such posturing reduces the likelihood of finding an optimal compromise. By contrast, the literature overwhelmingly concludes that information about policy outcomes improves the ability of citizens to hold governments to account.

Different mechanisms of information transmission – the media, FOILs, open decision-making processes, the collection of aggregate data – differ in their effectiveness in provid-
ing information about policy choices and outcomes. FOILs and open political institutions may improve the public’s ability to observe policy choices and the policy-making process. FOILs, for instance, allow members of the populace to legally demand information about the actions of and discussions within the government, thereby allowing the public to punish government actions taken against the (perceived) public interest. As noted by Prat (2005) and Stasavage (2004), access to this information may have both positive and negative welfare consequences – which effect dominates is theoretically ambiguous and likely to vary according to the degree to which the public is informed about the mapping between policy choices and consequences.

A free media with access to a wide audience may inform the public about both policy choices and policy outcomes. Investigative journalists may examine the manner in which government decisions are reached and may uncover policies the government attempts to hide. The media may also report on the outcomes of policy choices – with stories focusing on those most affected by a given set of government decisions. Two criteria must be satisfied for the media to effectively play this role: (1) news agencies must be free of government interference and (2) they must be able to reach large audiences.

Yet, the media are unlikely to be able to aggregate information on the broad welfare effects of policy decisions. Individual news agencies typically cannot invest the resources necessary to document movements in price indexes, levels of economic growth, levels of poverty, or education rates. The collection of such information often entails enormous fixed costs. Indeed, the collection of census data often involves workforces that number in the tens or hundreds of thousands. Moreover, such information has a public goods-like nature – once collected and reported by any one news agency, all other such agencies can simply reproduce these numbers. Aggregate data collection thus usually falls to the government. And the provision of such information is likely to affect the public’s awareness of the macro outcomes of policy decisions.
Attempts to measure transparency across states have generally focused on the role of the media, to the exclusion of other mechanisms of information dissemination. For instance, the commonly used Freedom House Freedom of the Press index measures subjective expert judgments of (1) the laws and regulations that constrain media content, (2) the degree of political control over the media, and (3) the structure of media ownership.

But, as noted above, the freedom of the media to conduct operations without government interference constitutes only one of at least two prerequisites for facilitating the flow of information. In addition to freedom of operations, these media must reach a sufficient portion of the population for their efforts to have a significant effect.

An alternative measure focuses on the level of newspaper circulation (see, for instance, Adserà, Boix and Payne, 2003). This index proxies for the ability of the press to reach a broad swath of the population and may be an important explanatory variable in some contexts. The newspaper circulation index, however, suffers from the opposite problem from that of Freedom House. High newspaper circulation numbers may be driven by the pervasiveness of state-run media outlets offering biased coverage of government actions.

Other measures of transparency tend to focus on the openness of the decision-making process of particular institutions. This is particularly true for the literature on central bank transparency. For instance, Stasavage (2003) constructs a measure based on the degree to which central banks publish economic forecasts (see also Chortareas, Stasavage and Sterne, 2002). These measures have the advantage of reflecting a particular form of transparency that is expected to affect outcomes in a precisely modeled manner. They do not reflect sufficiently broad indicators of government performance, however, to assess the effects of information on government accountability.

Scholars have paid relatively scant attention to the form of information dissemination discussed in this paper: the collection and reporting of aggregate economic data by the

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4See [http://freedomhouse.org/templace.cfm?page=533](http://freedomhouse.org/templace.cfm?page=533)
government. One notable exception is Islam (2006), who constructs an index based on the speed with which governments report data to the WDI, to the International Monetary Fund’s International Financial Statistics (IFS), and other sources. She examines 11 indicators drawn from across these series and assigns scores based on the degree of deviation from the ‘desirable’ frequency with which such data should be reported. These scores are then added together to construct an aggregate measure of data reporting frequency.

In the work that follows, we construct an index that, like Islam’s, focuses on data dissemination. Unlike Islam, we focus on the simple presence or absence of data from the WDI, rather than on the frequency with which data is reported. Our index (henceforth, the HRV Index) is also based on a large number of indicators and offers coverage for 149 countries over the full 1980-2007 period. We construct the index through an item response theory model that, as is outlined below, ensures minimal information loss from collapsing a 172 dimension observation into a single dimension representation.

We contend that the HRV Index is a valid measure of the degree of data dissemination by the government. Like other measures of transparency, ours is limited to a specific facet of information transmission. It thus has strengths and weaknesses, and its appropriateness will depend on the theory being tested. Our index, we argue, has one important property: that of aggregation. Aggregate data indicate the effectiveness of government policy better than citizens’ personal experiences, as we outline below.

2 The Importance of Aggregation

The release of economic data provides the public with information on policy outcomes – particularly macro level outcomes such as economic growth or inflation. Economic data are hardly unique in this regard. A free and vibrant media may also ease the flow of infor-

[Islam (2006)] also examines the role of an alternative form of transparency – the operation of FOILs.
mation about policy outcomes; reporters may examine the effects of government decisions as well as the decision-making process itself. Similarly, social networks may allow those impacted by policy decisions to communicate with one another and with those not directly targeted by policy choices.

Aggregate data are unique in one regard: their ability to provide information regarding a large number of individual outcomes. Newspaper stories are effective at communicating the association between policy choices and the welfare of individuals or small groups that may or may not be representative. In the absence of aggregate data, reporters cannot – by themselves – credibly convey the broader effects of policy choices. Moreover, the public goods property of information and the high fixed-costs of data collection ensure that even a vibrant media market will under-supply such information.

The collection and dissemination of aggregate data is likely to be critical to political accountability if three conditions hold: (1) The public is imperfectly informed as to the optimal policy choice. Were this not the case, one could know the appropriateness of the government’s actions simply by observing its policy decisions. A free and vibrant press, coupled with sufficiently open political institutions, would be vital to the transmission of such information; economic data would not.

(2) The policy in question affects a broad swath of the public. If policies are narrowly tailored, members of the media or social networks are likely able to discover the policy’s effects simply by contacting interested parties.

(3) Policy choices do not perfectly map into welfare implications for individuals. This

\footnote{We refer to optimality from the perspective of the majority, or of a member of the winning coalition. It is not necessary for all citizens to share policy preferences, though this assumption is often adopted for simplicity in retrospective voting models. For instance, assume that the utility of citizen \( i \) from a policy choice \( P \) is a function \( g(P) + \eta_i \) where \( \eta_i \) is a random shock that follows some commonly known distribution \( f(\cdot) \). Citizen uncertainty over the optimal policy would result if the citizens were uninformed of the functional form of \( g(P) \). Information regarding the choice of \( P \) would thus be of minimal value in securing government accountability.}

\footnote{This is equivalent to assuming that the distribution of \( \eta_i \), as discussed above, is not degenerate (e.g., \( \eta_i = 0 \forall i \)).}
requirement ensures that each individual’s welfare is an imperfect indicator of the ‘correctness’ of the government’s decisions. Thus, individuals should not optimally condition their decision of whether to discipline the government on their individual welfare alone.\footnote{See Kuran (1991) and Khemani (2007) for related discussions.} The weaker the correlation between government policies and individual outcomes, the more important aggregate information. The law of large numbers implies that data offer an increasingly precise signal of the appropriateness of government decisions as the level of aggregation rises.\footnote{This discussion relates to the literature on sociotropic versus ‘pocketbook’ voting in American politics (Kramer 1983; Markus 1988). Note further that the disclosure of more and more accurate economic data would provide greater information not only regarding the average effectiveness of government policies on individual welfare, but also more information regarding the distribution of these effects.}

We do not contend that the ‘law of large numbers effect’ fully solves the informational problems faced by the populace. Even if aggregate economic outcomes are fully known, assessing government responsibility for these outcomes is a difficult task. Economic well-being may fluctuate with a variety of shocks orthogonal to government policy. But, shocks to aggregate level measures of economic performance are the sum of a variety of similar shocks at the individual level. Our contention is that – in practical settings – the effects of such shocks on aggregate-level outcomes is likely to be smaller than those on individual-level outcomes, such that the provision of aggregate-level data enables citizens to draw better (though still imperfect) inferences about the appropriateness of government policies.

In addition to this ‘law of large numbers effect’, aggregation has implications for collective action. Collective action on the part of members of the populace is often inhibited by uncertainty over the willingness of others to participate in collective activities. This has particularly important implications for the accountability of autocrats, as methods of disciplining such governments – public protest, revolution – suffer from collective action problems. \footnote{Fearon (2006), for instance, finds that uncertainty over the distribution of dis-}
content may hinder public sanctioning of autocratic regimes. In his model, elections may serve as an informational device to overcome this problem. It is our contention that the public disclosure of aggregate data may do the same.

3 Missing Data as an Element of Transparency

Our concept of theoretical interest is the disclosure of aggregate economic information to the public. Our empirical measure of this concept is a function of the missingness/non-missingness of data from the WDI.

Few studies have employed data-availability as a variable, but those that do implicitly disagree as to the processes that drive this form of transparency. Bueno de Mesquita et al. (2003) and Hollyer, Rosendorff and Vreeland (2011) focus on missing data as a reflection of governments’ willingness to disseminate data. By contrast, Stone (2008) uses missing data as a proxy for state capacity, assuming that missing data result from the inability of a government to collect and publish information. We suggest that these two explanations are non-exclusive: both the government's willingness and its capacity to collect and distribute data are necessary, but neither is sufficient, to ensure transparency.

We suggest that transparency should not, in general, be viewed in contradistinction to state capacity. Nor does transparency only reflect the willingness (whether primitive or induced) of a government to provide information. In our view, transparency is a reflection of both the willingness of a government to provide information and its ability to do so. Governments with greater fiscal resources and/or a more technically competent bureaucracy are, ceteris paribus, more likely to be transparent. Conversely, governments with a stronger preference for disclosure – whether induced by political institutions (Broz, 2002; Hollyer, Rosendorff and Vreeland, 2011) or primitive – are likely to be more transparent. Those with a strong preference for secrecy will tend to avoid disclosure. For instance, Ross
(2006) notes that “high-performing authoritarian states” tend to report child mortality and economic data infrequently, despite relatively high *per capita* income levels and low mortality rates.

From the point of view of a citizen, it may not matter whether a paucity of data results from state willingness or capacity. In the absence of data disclosure, this citizen will be less able to draw accurate inferences about government behavior. Regardless of whether her government was unable or unwilling to collect and disclose data, she is equally uninformed.

This is not to say the question of whether willingness or capacity has a greater effect on disclosure is unimportant. To the contrary, this question is central to political science, and we offer a variable which scholars can use to test various hypotheses about which causal mechanism dominates under different circumstances.

Indeed, one can attribute the rankings of certain states on our index to either their capacity or their willingness to disclose data. For instance, Cuba receives the lowest transparency score of any country in our sample during the year 1980. At the time, Cuba's *per capita* GDP was nearly twice that of the median country in the Penn World Table, and the government had vastly expanded education provision to rates similar to those of the developed world (Barro and Lee 1996). Cuba possessed a high level of state capacity, but *chose* to not to disclose data to the WDI. By contrast, Somalia experienced precipitous drops in its HRV score in 1990 and 1991 – coinciding with the collapse of the Siad Barre government and the outbreak of civil war. It is probable that we can attribute most of the decline in transparency in this instance to the loss of state capacity, as Somalia became the paradigmatic example of a failed state.
4 The Empirical Model

We measure governments’ collection and disclosure of data directly, by relying on the presence or absence of reported values from the WDI. The World Bank assembles these data from information provided by other international organizations – for instance, the International Monetary Fund (IMF) and International Labor Organization (ILO). These organizations, in turn, obtain information from national statistical offices. In some instances, the World Bank codes observations in the WDI as missing because the information provided by national agencies is deemed to be questionable. Our measure thus reflects the disclosure of credible information by national statistical agencies. As such, it possesses a high degree of content validity (Carmines and Zeller, 1979) with regards to the concept of theoretical interest – the collection and dissemination of aggregate economic data.

To be more precise, we treat transparency as a latent (unobserved) term predicting the presence or missingness of data on 172 measures drawn from the WDI’s section pertaining to Economic Policy and Debt. We obtain estimates of this transparency term through the use of a Bayesian item response theory (IRT) model. IRT models are a class of procedures, drawn from the psychometrics literature (for a thorough overview, see van der Linden and Hambleton, 1997) and used to reduce the dimensionality of data with minimal loss of information. Such models have previously been used in political science to estimate the ideal points of legislators (Clinton, Jackman and Rivers, 2004), to refine indexes of democracy (Treier and Jackman, 2008; Pemstein, Meserve and Melton, 2010), to examine support for the US within the UN General Assembly (Voeten, 2004), and to combine expert estimates of bureaucratic agency ideologies (Clinton and Lewis, 2008).

Our approach assumes that there is an unobserved term – transparency – that drives the reporting of variables to the WDI. Countries that score highly on this term are more likely to report each item in the WDI than countries that receive low scores; though the degree
to which reporting/non-reporting reflects transparency varies across each variable in the WDI. There is likely to be variation in the degree to which reporting reflects transparency both because certain variables are more likely to be reported – by all countries – than others, and because certain variables may be reported for idiosyncratic reasons. Thus, the reporting/non-reporting of different variables in the WDI provides differing amounts of information about transparency. Our model estimates the degree to which each observation is informative. We can therefore report a single value of transparency for each country-year that reflects the missingness of data across all 172 WDI variables with minimal loss of information.

In our model, we let \( y_{j,c,t} \in \{0,1\} \) denote an indicator equal to 1 if country \( c \) reports WDI variable \( j \) in year \( t \) and equal to 0 otherwise. We then estimate

\[
Pr(y_{j,c,t} = 1|\text{transparency}_{c,t}) = \logit(\delta_j + \beta_j \text{transparency}_{c,t})
\]

where \( \delta_j \) is the difficulty parameter and \( \beta_j \) the discrimination parameter for item \( j \). The term \( \text{transparency}_{c,t} \) is the measure of a given country-year’s propensity to disclose data, which is to be estimated. The \( \logit \) function is a logistic transformation. Thus, changes in \( \delta_j \) reflect the degree to which countries, on average, report a given variable drawn from the WDI. Changes to \( \beta_j \) reflect the degree to which the outcome of one item predicts the outcome of other items (Gelman and Hill, 2006).

In our model there are 172 items \( j \), 149 countries \( c \) and 28 years \( t \). We thus estimate a system of 172 equations (all of the form of equation 1) with 4172 observations.\(^{10}\)

The model represented by equation 1 is only identified up to an affine transformation.

\(^{10}\)We confine our analysis to 149 countries that have been continuously independent since 1980 and are recorded in the WDI from that time. We rely on the DD Index of Democracy (Cheibub, Gandhi and Vreeland, 2010) to define independent country years. The WDI provide data for dates preceding independence. But, we do not rely on such data, given that these must either be reported by entities other than that of the sovereign government or are interpolated from information reported subsequently. We thus run our model on a rectangular dataset.
The analyst never observes the actual values of $\delta_j$, $\beta_j$ or $\text{transparency}_{c,t}$. One could therefore multiply all transparency values by a constant, and divide all $\beta_j$ by the same value, and produce the same model. One could similarly manipulate coefficient values to shift the transparency index in any direction. To avoid this problem, it is standard to treat the values of transparency as draws from an informative prior distribution. By so anchoring the values of transparency, the difficulty and discrimination parameters become estimable (Gelman and Hill, 2006; Clinton, Jackman and Rivers, 2004).

In this instance, however, we encounter an additional modeling difficulty. Since our data are time-series cross-sectional, we cannot treat each observation of $\text{transparency}_{c,t}$ as independent of the observation $\text{transparency}_{c,t-1}$. To adjust for the likely correlation of index values within countries over time, we rely on a system of dynamic priors. We treat our prior beliefs of the transparency score for each country in each year as a function of the transparency score for that country in the previous year. This set of priors will ‘shrink’ the estimate of a given country-year’s transparency score back towards the estimated level of transparency for that country in the previous year, smoothing discrete movements in our index. Our system of priors is thus: $\text{transparency}_{c,t} \sim N(\text{transparency}_{c,t-1}, \frac{1}{\tau_c}) \forall t > 1$ and $\text{transparency}_{c,t} \sim N(0, 2)$ for $t = 1$. $\tau_c$ acts as a country specific smoothing parameter, since the degree to which an estimated value of $\text{transparency}_{c,t}$ shrinks back towards the prior mean is inversely proportional to the variance of the prior distribution (Jackman, 2009). Similar priors are often used in the ideal point estimation literature when working with dynamic data (see, for instance Martin and Quinn, 2002). We estimate the parameter $\tau_c$ for each country in our data, and give this term a prior $\tau_c \sim \text{Gamma}(40, 0.1)$.[11]

We place diffuse normal priors on the difficulty and discrimination parameters $\delta_j$ and

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[11] A Gamma distribution with a scale parameter 40 and shape parameter 0.1 has a mean of 4 and a variance of 0.4.
\( \beta_j \), such that\(^{12} \)

\[
\begin{pmatrix}
\delta_j \\
\beta_j
\end{pmatrix}
\sim N\left( \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 100 & 0 \\ 0 & 100 \end{pmatrix} \right).
\]

These diffuse priors ensure that the posterior estimate of each discrimination and difficulty parameter is overwhelmingly determined by the data.

We estimate this model employing a Markov Chain Monte Carlo algorithm run from WinBUGS 1.4.3 launched from R.2.11.0. The model is estimated using 2 chains of 250,000 iterations each, the first 150,000 of which serve as a burn-in period.

5 Influential Items

Some reasonable concern remains as to whether the index derived from the model above reflects the dissemination of \textit{politically relevant} data. We construct our index based on the reporting or non-reporting of 172 variables, some of which are obscure and of little value to the public in disciplining government behavior. If these items drive most or all of the variation in the HRV Index, our measure is a poor barometer of transparency and is unlikely to affect government accountability.

Fortunately, we can assess whether politically relevant variables influence the HRV Index scores by examining the posterior densities of the difficulty (\( \delta_j \)) and discrimination (\( \beta_j \)) parameters described in equation 1. Recall that the discrimination parameters (\( \beta_j \)) reveal the extent to which the reporting/non-reporting of a given variable correlates with the transparency index; while the difficulty parameters (\( \delta_j \)) reflect the extent to which a

\(^{12}\)By treating these priors as independent draws from a multivariate normal, rather than draws from two univariate normals, we ‘block’ on these parameters. Blocking implies that the Gibbs sampler simultaneously draws simulated values of both \( \delta_j \) and \( \beta_j \), rather than conditioning the draw of each term on prior draws of the other. Blocking on parameters that are likely to be highly correlated speeds the rate at which the Gibbs sampler explores the parameter space (Jackman 2009).
given observation is reported in the data. Large and positive values of the $\beta_j$ indicate that a given variable is much more likely to be reported in a highly transparent country, a negative $\beta_j$ indicates the reverse, and values close to zero indicate that the reporting of a given variable does not much affect our transparency index.

The variables whose dissemination has the strongest influence on the HRV Index are highly politically relevant. The most discriminating items include various measures of household expenditures, government expenditures, gross national expenditures, exports, and imports. Highly politically relevant variables such as GDP and GDP per capita have large and positive discrimination parameter estimates (generally in the vicinity of 10-12). Measures whose reporting negatively correlates with index values are uniformly measures of development assistance, which are not typically reported by developed countries.

Those variables whose reporting exercises little to no influence over the HRV Index include such measures as workers’ remittances, changes in inventories, manufacturing value added, and the net capital account. Typically these data would not be expected to be highly politically influential. Figures 1 and 2 plot the discrimination and difficulty parameter estimates, and 95 percent highest posterior density intervals, for the 25 items with the highest and lowest discrimination parameter estimates, respectively.\footnote{The abbreviation ‘LCU’ stands for ‘Local Currency Unit’, while ‘DAC’ stands for ‘Development Assistance Committee’.}

\section{Face Validity}

To assess the validity of our estimates, we first examine the estimated values of the transparency index in the 1980 cross-section. We report estimated index values and 95 percent highest posterior density intervals for the 25 lowest scoring countries in Figure 3 and for the 25 highest scoring countries in Figure 4.\footnote{Values from the MCMC estimates were initially inverted such that low scores corresponded to higher levels of transparency. We have rescaled the index by multiplying all values by $-1$. We similarly rescale all}
Discrimination and difficulty parameters, and 95 percent highest posterior density intervals, for the 25 items with the highest discrimination parameter scores. Parameter values are plotted on the x-axis. Item names are noted on the y-axis.

All the highest scoring countries are members of the OECD. While the lowest scoring countries consist of Cuba, Cambodia, Djibouti, Angola and Laos – all states we would expect to score poorly on transparency during this period. We can also see that the highest scoring country in this period (Austria) differs significantly in its score from other developed countries such as Denmark and the USA. Similarly, the lowest scoring country (Cuba) differs discrimination parameters.
Discrimination and difficulty parameters, and 95 percent highest posterior density intervals, for the 25 items with the lowest discrimination parameter scores. Parameter values are plotted on the x-axis. Item names are noted on the y-axis.

Significantly from other low scoring countries such as Equatorial Guinea and Qatar. Not only do our index values coincide with common notions of transparency, but we are also able to discriminate among high and low scoring countries. It is more difficult, however, to discriminate among the large number of countries that are placed in the middle ranges of the index (between 0 and 0.5), though the low and high ends of this range are statistically distinguishable.
A cross section of the 50 lowest scoring countries on the HRV Index from 1980. Three letter ISO country codes are plotted on the y-axis. Index scores are plotted on the x-axis. Mean predicted transparency index values are indicated by diamonds, while the whiskers denote 95 percent highest posterior density intervals. Countries are plotted in ascending order of mean HRV Index score.

In addition to comparing across countries cross-sectionally, we examine longitudinal variation in our estimates to assess their validity. Based on anecdotal accounts, the level of transparency in many countries in our sample should be shifting over time. We present our estimates of the level of transparency in several such countries below.

For instance, Vietnam experienced enormous economic and political changes over the 1980-2007 period. These changes included substantial economic liberalization, the creation of (relatively) more representative institutions, and the establishment of diplomatic relations with nations with which it had been estranged. The 1986 Party Congress is widely
A cross section of the 50 highest scoring countries on the HRV Index from 1980. Three letter ISO country codes are plotted on the y-axis. Index scores are plotted on the x-axis. Mean predicted transparency index values are indicated by diamonds, while the whiskers denote 95 percent highest posterior density intervals. Countries are plotted in ascending order of mean HRV Index score.

seen as ushering in the period of liberalization (Fforde and De Vylder, 1996; Riedel and Turley, 1999). The process of economic and political opening continued throughout the 1990s and early 2000s (Malesky, 2008, 2009; Gehlbach and Malesky, forthcoming). One might therefore expect that Vietnam would grow substantially more transparent over time, particularly after 1986.

Figure 5 plots our predicted transparency index scores and 95 percent highest posterior density intervals for Vietnam over time. As is visible from the figure, predicted levels of transparency increase substantially and significantly over time. There are large jumps in
the 1984-1986 period, corresponding to the Party Congress, such that the transparency score is significantly higher in 1986 than it was in 1985 and significantly higher in 1985 than 1984. Similar increases in transparency coincide with episodes of political and economic liberalization in Tanzania and China (see supporting materials).

Figure 5: Vietnam HRV Scores Over Time

A longitudinal plot of Vietnam’s HRV Index scores from 1980-2008. HRV Index scores are on the y-axis, while time is measured on the x-axis. Diamonds denote mean predicted index scores. Whiskers denote 95 percent highest posterior density intervals.

We do not observe a universal trend towards rising transparency levels over the period between 1980 and 2008 – a number of countries experienced substantial declines. For instance, Zimbabwe experienced increasing levels of unrest and trade union-led opposition to the ZANU-PF government beginning in the mid-1990s. General strikes were held in 1996, and met with police intimidation (van der Walt 1998). This period also saw the forced confiscation of white-owned farms setting off a precipitous economic decline and
hyperinflation in the early and mid-2000s. Levels of political repression increased dramatically prior to the 2008 elections pitting the incumbent Robert Mugabe against a strong opposition lead by Morgan Tsvangirai and Simba Makoni. The Mugabe regime began to plan – and to implement – its repressive strategy in 2007. This period saw dramatic clashes between pro-regime forces and opposition figures, including the beating of Tsvangirai by riot police (Meredith, 2007, 238-239).

Figure 6 plots the HRV Index scores for Zimbabwe over time. Zimbabwe’s transparency level falls sharply in 1995 – during the initial period of labor strikes and protests. Another substantial drop in transparency is evident in 2007 – one year before the hotly contested presidential elections that saw the ZANU-PF party rely on repressive tactics to maintain power. Our measure of transparency declines nearly in lock-step with this period of increased repression. Still more dramatic falls in transparency can be seen in episodes of state failure – for instance, following the outbreak of civil war in Somalia in 1990 (see supporting materials).

These results support the validity of the HRV Index as a measure of transparency – or, more precisely, the dimension of transparency associated with the collection and dissemination of economic data by the government. First, this index is built on a direct measure of data collection, the publication of information in the WDI. Second, the resultant index scores produce cross-sectional estimates that correlate highly with commonly held notions of transparency and that discriminate well amongst countries, particularly at the upper and lower end of the transparency spectrum. Finally, the HRV Index scores seem to move in line with developments within countries over time that one would expect to affect the degree of this form of transparency. Index scores improve dramatically with economic and political liberalization. And they decrease sharply with levels of repression and unrest.

---

15 For an overview of political developments in Zimbabwe during the mid- and late-1990s, see Sithole (2001). On the collapse of business-state relations see Taylor (2007, 30-34).
7 The Relationship between Government Data Collection and Other Forms of Transparency

We have established that the HRV Index moves in line with perceived levels of transparency – in both cross-sectional and longitudinal settings. It remains to be seen how this index correlates with commonly used indexes of transparency, indexes that attempt to measure alternative dimensions of the flow of information. *Ex ante*, one might expect these indexes to be positively correlated. Governments that are willing to allow freedom of the press or uninhibited access to the news media are more likely to encourage the free flow of information more generally. A free and active press may also actively push the government
to collect aggregate data and make this information publicly available.

Yet, one might also expect systematic differences between the HRV Index and such measures as Freedom House’s Freedom of the Press Index and daily newspaper circulation figures. Very little state activity is required to enforce the ‘negative rights’ implied by freedom of the press. By contrast, the collection and dissemination of aggregate economic data is often a costly exercise. The Freedom House measures are also related to democratic notions of individual rights. While it may be the case that democracies have a stronger incentive to collect and disseminate data than autocracies (Hollyer, Rosendorff and Vreeland, 2011), these incentives are less clearly associated with democracy than are protections of the freedom of the press.

Table 1 explores the relationship between the Freedom House and HRV Indexes. (We present scatterplots of the relationship between these two indexes in the appendix.) HRV Index scores are regressed on the Freedom House Index, on an indicator for democracy (we use the DD/ACLP definition of democracy as coded by Cheibub, Gandhi and Vreeland, 2010), GDP per capita in thousands of constant PPP adjusted 2005 US dollars (Heston, Summers and Aten, 2009), and the interaction between these latter two terms.

These results indicate that the two indexes are positively correlated. The coefficients on democracy, income and their interaction indicate that low and middle income (incomes below roughly $3,900) autocracies score highly on the HRV measure relative to the Freedom House measure. At higher incomes, democracies perform more strongly on the HRV Index than on that composed by Freedom House. Further investigations suggests that the relatively high scores of OECD countries on the HRV Index drive this relationship.

---

16 For instance, the most recent Indian census required 2.7 million workers, who often had to travel enormous distances, to complete the enumeration. “Heads Up: A National Head Count Should Show Dramatic Changes.” The Economist. February 24, 2011. http://www.economist.com/node/18233732?story_id=18233732

17 We weight all observations in these regressions by the inverse of the standard deviation of the HRV Index predictions.
Table 1: HRV Index Regressed on Freedom House Scores

<table>
<thead>
<tr>
<th></th>
<th>HRV Index</th>
<th>HRV Index</th>
<th>HRV Index</th>
<th>HRV Index</th>
</tr>
</thead>
<tbody>
<tr>
<td>FH Index</td>
<td>0.119***</td>
<td>0.113***</td>
<td>0.082***</td>
<td>0.024</td>
</tr>
<tr>
<td></td>
<td>[0.072,0.167]</td>
<td>[0.058,0.167]</td>
<td>[0.029,0.136]</td>
<td>[-0.031,0.080]</td>
</tr>
<tr>
<td>Democracy</td>
<td>0.020</td>
<td>0.041</td>
<td>-0.086**</td>
<td></td>
</tr>
<tr>
<td></td>
<td>[-0.058,0.098]</td>
<td>[-0.033,0.114]</td>
<td>[-0.171,-0.000]</td>
<td></td>
</tr>
<tr>
<td>GDP per capita</td>
<td></td>
<td>0.007**</td>
<td>-0.002</td>
<td>0.022***</td>
</tr>
<tr>
<td></td>
<td></td>
<td>[0.001,0.014]</td>
<td>[-0.004,0.001]</td>
<td></td>
</tr>
<tr>
<td>Democracy × GDP per capita</td>
<td>-0.271***</td>
<td>-0.248***</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>[-0.343,-0.198]</td>
<td>[-0.310,-0.187]</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Constant</td>
<td>-0.179***</td>
<td>-0.190***</td>
<td>-0.271***</td>
<td>-0.248***</td>
</tr>
<tr>
<td></td>
<td>[-0.222,-0.136]</td>
<td>[-0.253,-0.127]</td>
<td>[-0.343,-0.198]</td>
<td>[-0.310,-0.187]</td>
</tr>
<tr>
<td>(\hat{\sigma})</td>
<td>0.341</td>
<td>0.341</td>
<td>0.332</td>
<td>0.312</td>
</tr>
<tr>
<td>N</td>
<td>1637</td>
<td>1637</td>
<td>1637</td>
<td>1637</td>
</tr>
</tbody>
</table>

Results of a regression of HRV Index scores against the Freedom House Index, a democracy indicator, GDP *per capita* (in thousands of PPP US $), and the interaction of GDP *per capita* and democracy. 95 percent confidence intervals are presented in parentheses. All standard errors are clustered by country. \(\ast\ast\ast\) denotes significance at the 99 percent level, \(\ast\ast\) denotes significance at the 95 percent level, and \(\ast\) denotes significance at the 90 percent level. \(\hat{\sigma}\) represents the standard error of the regression.

Table 2 presents the results of a regression of HRV Index values against newspaper circulation numbers.\(^1\)\(^8\) (We also present scatterplots comparing these two indexes in the appendix.) The two indexes are clearly strongly related to one another, even controlling for additional covariates. There are systematic differences, however. In particular, democracies – and especially high income democracies – are ranked as more transparent by the HRV Index than by daily newspaper circulation. High income autocracies, by contrast, are ranked as significantly less transparent by the HRV Index. This finding suggests that high newspaper circulation numbers may be driven by a pervasive state media in autocracies. This result is important since our theoretical expectations about the role of transparency

\(^1\)\(^8\)All observations have been weighted by the inverse of the standard deviation of the HRV Index estimates.
relate not simply to the volume but to the quality of information relayed to the public. It seems that the HRV Index does a better job of reflecting this quality.

Table 2: HRV Index Regressed on Daily Newspaper Circulation

<table>
<thead>
<tr>
<th></th>
<th>HRV Index</th>
<th>HRV Index</th>
<th>HRV Index</th>
<th>HRV Index</th>
</tr>
</thead>
<tbody>
<tr>
<td>News Circ.</td>
<td>0.003***</td>
<td>0.002***</td>
<td>0.002*</td>
<td>0.002**</td>
</tr>
<tr>
<td></td>
<td>[0.001,0.004]</td>
<td>[0.001,0.004]</td>
<td>[-0.000,0.003]</td>
<td>[0.000,0.004]</td>
</tr>
<tr>
<td>Democracy</td>
<td>0.200*</td>
<td>0.186</td>
<td>-0.119</td>
<td></td>
</tr>
<tr>
<td></td>
<td>[-0.021,0.422]</td>
<td>[-0.045,0.416]</td>
<td>[-0.274,0.037]</td>
<td></td>
</tr>
<tr>
<td>GDP per capita</td>
<td>0.008</td>
<td></td>
<td>-0.023**</td>
<td></td>
</tr>
<tr>
<td></td>
<td>[-0.007,0.024]</td>
<td>[-0.041,-0.005]</td>
<td>[-0.041,-0.005]</td>
<td></td>
</tr>
<tr>
<td>Democracy × GDP per capita</td>
<td>0.036***</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>[0.022,0.050]</td>
</tr>
<tr>
<td>Constant</td>
<td>0.169***</td>
<td>0.078</td>
<td>0.057</td>
<td>0.246***</td>
</tr>
<tr>
<td></td>
<td>[0.090,0.247]</td>
<td>[-0.027,0.183]</td>
<td>[-0.050,0.165]</td>
<td>[0.161,0.331]</td>
</tr>
<tr>
<td>σ</td>
<td>0.515</td>
<td>0.506</td>
<td>0.503</td>
<td>0.473</td>
</tr>
<tr>
<td>N</td>
<td>476</td>
<td>476</td>
<td>476</td>
<td>476</td>
</tr>
</tbody>
</table>

Results of a regression of HRV Index scores against the World Bank’s Daily Newspaper circulation per 1000 numbers, a democracy indicator, GDP per capita (in thousands of PPP US $), and the interaction of GDP per capita and democracy. 95 percent confidence intervals are presented in parentheses. All standard errors are clustered by country. *** denotes significance at the 99 percent level, ** denotes significance at the 95 percent level, and * denotes significance at the 90 percent level. σ represents the standard error of the regression.

So, while the HRV Index correlates with other measures of transparency, it also exhibits systematic differences. Poor- and middle-income autocracies appear more willing to disseminate data than to tolerate freedom of the press. The contrast between high-income democracies and other countries is stronger with respect to data dissemination than other forms of transparency. In the next section, we examine whether these systematic differences across dimensions of transparency are associated with variation in the association between these measures and governance.
8 Transparency and the Quality of Governance

Political scientists have long been concerned with political accountability and its effects on government performance. A seminal contribution to this literature is that of Adserà, Boix and Payne (2003) (henceforth, ABP), which examines the effect of democracy and the free flow of information on government corruption, bureaucratic quality, and the rule of law. Democracies, they find, perform significantly better than autocracies in all four measures – and the role of democracy is further accentuated when information (proxied by the level of circulation of daily newspapers) flows freely.

We replicate the work of ABP below, using both the HRV Index of data dissemination and the circulation of daily newspapers as measures of different forms of transparency. Our theoretical expectations, based on the discussion above, lead us to believe that (1) the HRV Index is likely to be more predictive of government performance in autocratic systems – where collective actions by citizens are necessary to discipline the government, and (2) the HRV Index is likely to be less predictive of government behavior than media activities when the citizenry is likely to be certain of its preferred policy choice. Note that the majority of governance indicators examined here are likely to involve relatively low uncertainty over policy choices, so this is a conservative test of the role of data dissemination on government performance.

19 ABP also consider the relationship with the risk of appropriation. The current version of the ICRG dataset, however, does not include this measure.

20 Ours is not an exact replication of ABP, as several of the datasets they use have been updated and modified. We use the most recent ICRG measures of corruption, law and order, and bureaucratic quality. The bureaucratic quality measure in our specifications ranges from 0-4, all other measures range from 0-6. ABP measure democracy using the Polity III democracy measure. We rely on the more commonly used polity2 measure drawn from the Polity IV dataset. Finally, we append additional observations of daily newspaper circulation through 2004 (drawn from the 2011 WDI). Our specification also differs from that of ABP. They include an interaction between newspaper circulation and democracy in their model, but assume that newspaper circulation has no direct effect on governance. We prefer not to make such an assumption and, following the advice of Brambor, Clark and Golder (2005), include the constitutive terms of all interactions in our specifications.

21 We do not intend for these results to be interpreted causally. As noted above, our transparency scores are likely to be correlated with state capacity, which may act as an omitted variable driving the relationship
We follow ABP in treating each governance indicator as a linear function of democracy (here proxied by the polity2 score), GDP per capita measured in 2005 constant purchasing power parity US dollars, transparency, the interaction of transparency and democracy, and a lagged dependent variable. The model is thus

\[ \text{governance}_{i,t} = \alpha + \phi \text{governance}_{i,t-1} + \gamma_1 \text{democracy}_{i,t} + \gamma_2 \text{transparency}_{i,t} + \gamma_3 \text{democracy}_{i,t} \times \text{transparency}_{i,t} + \gamma_4 \text{GDP}_{i,t} + \epsilon_{i,t}. \]  

(2)

To improve the comparability of regression coefficients, we standardize both the HRV Index and the newspaper circulation values by subtracting the mean from each observation and dividing by the standard deviation. The coefficients on these terms thus measure the contemporaneous association between a one standard deviation shift in the transparency measure and governance. (The presence of a lagged dependent variable makes the model dynamic. Thus, the steady state association between these terms and governance is given by dividing each coefficient value by \(1 - \phi\).)

Also following ABP, we collapse yearly observations into half-decade intervals, taking the mean of all terms except newspaper circulation in each five year period. As newspaper circulation is only measured in five year increments prior to 1997, we simply take the newspaper circulation figures from the first year in each period as the observation. Collapsing the data helps to adjust for slow movement in the governance indicators over time and helps to reduce measurement error in these terms.

Lagrange multiplier tests of the model depicted in equation 2 under the null hypothesis that the disturbances are not serially correlated reject this hypothesis, even after averaging with governance outcomes. This problem is likely to be particularly severe with respect to measures of bureaucratic quality. While controlling for GDP per capita helps to adjust for this problem, it may not solve it. We note that such bias is unlikely to explain why the HRV Index is more strongly associated with governance outcomes in autocratic – as opposed to democratic – states. Our purpose here is to ascertain the validity of our measure by replicating a seminal work in the literature.

30
ing over five year periods and the inclusion of a lagged dependent variable. We therefore follow the advice of Beck and Katz (1995) and Keele and Kelly (2006) and adjust for residual first order autocorrelation by estimating equation 2 using the Prais-Winston GLS method. All models are estimated with Huber-White corrected standard errors to adjust for heteroscedasticity, which may be induced by measurement error in the HRV and newspaper circulation indexes. Table 3 presents the results of these regressions.

To ease in the interpretation of the results of these regressions, we plot the marginal contemporaneous marginal effect of a change in each transparency measure for each possible value of polity in Figures 7, 8, and 9 (Brambor, Clark and Golder, 2005). In each graph, the values of the polity score are plotted on the x-axis, and the values of the marginal effect are plotted on the y-axis. 95 percent confidence intervals are denoted by dashed lines.

Data dissemination is more strongly associated with the quality of governance – across all measures – in autocracies than in democracies. By contrast, the role of newspaper circulation varies only slightly with regime-type. Typically, the precision of the estimates of the marginal effect of newspaper circulation increases in the polity2 score, such that the marginal effects of newspaper circulation are only significantly different from zero in democratic states.

This result is in keeping with theoretical expectations. Recall that data dissemination may play a particularly strong role in facilitating collective action to discipline govern-

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22 One might also estimate an autoregressive distributed lag or error correction model – which would treat the serial correlation of the error term as the subject of theoretical interest rather than a nuisance to be discarded (Beck and Katz forthcoming, 2011). However, to maintain maximum comparability with ABP, we prefer here to present the GLS corrections.

23 An alternative would be to weight each observation by the inverse of the standard deviation of the HRV Index estimate. Unfortunately, however, the Prais-Winsten estimation method precludes the use of regression weights in Stata.

24 These plots are for models that control for both the HRV Index and newspaper circulation. We stress that we mean the term ‘marginal effect’ simply in the sense of the first partial derivative with respect to a given measure of transparency. We do not wish to suggest that we are inferring causality based on these results.
Plots of the marginal effect of transparency on bureaucratic quality for varying levels of democracy. Polity2 (democracy) scores are plotted on the x-axis, the marginal effect of a one-standard deviation change in each transparency measure is plotted on the y-axis. Point estimates are represented by the solid line, 95 percent confidence intervals are represented by the dashed lines.

Data about aggregate outcomes help to inform members of the public about general outcomes, and thus help them better predict the actions of their fellow citizens. As members of the public cannot discipline their government through the polls in an autocracy, they must rely on the threat of collective action to remove or otherwise punish the government. Thus, one would expect data dissemination to be particularly important in autocracies. Of course, these results are hardly proof of this mechanism. We cannot imply causality from these results, and attempts to do so are well beyond the scope of this paper. Nonetheless, this finding is suggestive, and the contrast between the different facets of transparency is clear.
Figure 8: Marginal Effects of a Standard Deviation Change in Transparency on Law and Order

Plots of the marginal effect of transparency on law and order for varying levels of democracy. Polity2 (democracy) scores are plotted on the x-axis, the marginal effect of a one-standard deviation change in each transparency measure is plotted on the y-axis. Point estimates are represented by the solid line, 95 percent confidence intervals are represented by the dashed lines.

Across all measures of governance, the HRV Index is a stronger predictor of government performance than is newspaper circulation in all but the most democratic countries. The point estimates on the contemporaneous marginal effect of the HRV Index suggest that a one standard deviation shift in index scores is associated with a 0.65 to 0.75 standard deviation change in governance indexes in the most autocratic nations, and that this effect declines as governments grow more democratic. By contrast, a one standard deviation increase in newspaper circulation is associated with a 0-0.2 standard deviation improvement in governance indexes in autocratic governments, and a broadly similar change in
The comparatively large magnitude of the association between data dissemination and governance quality is somewhat surprising for some of these measures. Given that the public is likely to be well-informed of optimal policies with respect to corruption and law and order, one might expect a free press to play a larger role in determining the quality of governance along these dimensions. The measurement schemes employed by the ICRG may play a roll in this finding. Their measure of law and order encompasses not just such elements as the independence and fairness of the judiciary, but also measures of the

\footnote{Note that the lagged dependent variable coefficients indicate that the steady-state effects of such shifts are 1.5 to 2 times this size.}
level of criminality. As the latter may be influenced by the appropriateness of government policing strategies – the optimal choice of which are unlikely to be known ex ante by the public – data dissemination may play a role in improving policy choices. Similarly, the ICRG’s corruption measure refers to the extent to which the spheres of government and business are intertwined, rather than specifically to acts of bribery or misappropriation by civil servants. Here too, data dissemination may improve the public’s ability to sanction policy choices.

Our results thus demonstrate that the HRV Index is robustly correlated with governance outcomes, controlling for an alternative form of transparency (i.e., newspaper circulation). For most countries, this association is stronger – by several orders of magnitude – than that between newspaper circulation and governance. These findings are broadly consistent with theoretical expectations.
Table 3: Governance Measures Regressed on Transparency

<table>
<thead>
<tr>
<th></th>
<th>Bureaucratic Quality</th>
<th>Law and Order</th>
<th>Corruption</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lagged DV</td>
<td>0.685***</td>
<td>0.670***</td>
<td>0.646***</td>
</tr>
<tr>
<td></td>
<td>[0.628,0.742]</td>
<td>[0.611,0.730]</td>
<td>[0.565,0.727]</td>
</tr>
<tr>
<td>Polity 2</td>
<td>0.018***</td>
<td>-0.017</td>
<td>0.016*</td>
</tr>
<tr>
<td></td>
<td>[0.008,0.028]</td>
<td>[-0.021,0.008]</td>
<td>[0.005,0.028]</td>
</tr>
<tr>
<td>Newspaper Circ.</td>
<td>0.027</td>
<td>0.141</td>
<td>0.138</td>
</tr>
<tr>
<td></td>
<td>[-0.096,0.150]</td>
<td>[-0.002,0.285]</td>
<td>[0.053,0.395]</td>
</tr>
<tr>
<td>Newspaper Circ. × Polity 2</td>
<td>0.005</td>
<td>-0.001</td>
<td>0.002</td>
</tr>
<tr>
<td></td>
<td>[-0.007,0.018]</td>
<td>[-0.014,0.014]</td>
<td>[-0.015,0.018]</td>
</tr>
<tr>
<td>GDP per capita</td>
<td>0.017***</td>
<td>0.025***</td>
<td>0.010***</td>
</tr>
<tr>
<td></td>
<td>[0.011,0.023]</td>
<td>[0.013,0.036]</td>
<td>[0.005,0.015]</td>
</tr>
<tr>
<td>HRV Index</td>
<td>0.273*</td>
<td>0.453*</td>
<td>0.346*</td>
</tr>
<tr>
<td></td>
<td>[0.018,0.527]</td>
<td>[0.088,0.819]</td>
<td>[0.068,0.625]</td>
</tr>
<tr>
<td>HRV Index × Polity 2</td>
<td>-0.026*</td>
<td>-0.039*</td>
<td>-0.029*</td>
</tr>
<tr>
<td></td>
<td>[-0.052,-0.000]</td>
<td>[-0.076,-0.002]</td>
<td>[-0.057,-0.000]</td>
</tr>
<tr>
<td>Constant</td>
<td>0.495***</td>
<td>1.530***</td>
<td>0.975***</td>
</tr>
<tr>
<td></td>
<td>[0.349,0.641]</td>
<td>[1.284,1.777]</td>
<td>[0.656,1.295]</td>
</tr>
<tr>
<td>σ</td>
<td>0.423</td>
<td>0.710</td>
<td>0.629</td>
</tr>
<tr>
<td></td>
<td>0.384</td>
<td>0.726</td>
<td>0.629</td>
</tr>
<tr>
<td>ρ</td>
<td>0.236</td>
<td>0.075</td>
<td>0.074</td>
</tr>
<tr>
<td>N</td>
<td>283</td>
<td>283</td>
<td>283</td>
</tr>
</tbody>
</table>

Results of a regression of ICRG governance indicators against differing measures of transparency. Both the HRV Index and the newspaper circulation values have been standardized. Estimates are obtained through GLS using the Prais-Winsten correction for first order autocorrelation. All standard errors have been clustered at the country level. 95 percent confidence intervals are presented in brackets. * denotes significance at the 90 percent level, ** denotes significance at the 95 percent level, and *** denotes significance at the 99 percent level.
9 Conclusion

We have argued that transparency can be decomposed along various dimensions – the freedom of the media and the functioning of media markets, institutional forms that facilitate accountability and public access, and the voluntary dissemination of data by the government. These varying forms of transparency operate according to different mechanisms and should be expected to have differing effects on government behavior. We have argued that a crucial feature of government data dissemination lies in its ability to communicate information about aggregate level outcomes to members of the citizenry, allowing citizens to adopt more effective strategies to discipline the government when optimal policy choices are uncertain – and that these effects are likely to be particularly important under autocratic rule.

We have further constructed an index measuring data dissemination with many desirable properties. Our index is based on objective criteria, rather than subjective judgments, and reflects the dissemination of credible information. The HRV also provides information on a far wider range of countries and dates than alternative transparency measures. Moreover, it captures a largely unmeasured form of transparency of great theoretical import.

In addition to constructing this index, we show that (1) it systematically differs from other commonly used indexes in political science, and (2) that these differences have empirical meaning. The HRV Index has a positive association with governance outcomes even after alternative forms of transparency are controlled for – and this association is substantially stronger with our index than with alternate facets of transparency. It is particularly strongly associated with good governance in autocracies.

Of course, much remains to be done to more precisely document the mechanisms by which data dissemination affects government decision-making. Scholars should take greater care in specifying theoretical mechanisms, in constructing outcome measures, and
in documenting the causal influence of transparency. Indeed, one direction of future work should reflect the decision of governments to release such information (for one example in this vein, see Hollyer, Rosendorff and Vreeland [2011]). This paper takes a necessary step in this direction by providing a valid measure of data dissemination.

References

Adserà, Àlicia, Carles Boix and Mark Payne. 2003. “Are You Being Served? Political Ac-


Series-Cross-Section Political Economy Data.” *Annual Review of Political Science* .


A Appendix

A.1 The Relationship Between Freedom of the Press and the HRV Index

Figure 10 depicts a scatterplot of HRV Index scores against Freedom House scores both from 1994-2004. Both indexes have been standardized by subtracting the mean index value from each observation and dividing by the standard deviation so that each index is on the same scale with mean zero and standard deviation 1. The dashed line follows a 45 degree angle, so the two measures would be perfectly correlated if all measures fell on this line. Observations that fall above the line score relatively highly on the Freedom House Index (plotted on the y-axis) as compared to the HRV Index (plotted on the x-axis). The reverse holds for observations that fall below the line. The dotted curve represents a locally weighted scatterplot smoother (lowess) that is applied to these data.

The correlation between the two indexes is apparent in the figure. The slope of lowess is close to one for much of the range of both indexes; though it flattens to levels close to zero at very high levels of both scores. The two indices broadly agree on which countries should receive relatively high and relatively low transparency scores – though the magnitude of the difference between high/low-scoring countries and mean index values is far greater in the HRV Index than in Freedom House. This relationship may reflect the constrained range of values (0-100) that the Freedom House Index can assume. Freedom House, however, assigns a wide range of scores to country-years that rank in the middle of the HRV Index. Perhaps this relationship reflects differences in the way the two indices evaluate laws and institutions versus practices. Roughly one-third of the Freedom House index is determined by questions relating to legal protections for freedom of speech.

26 The Freedom House Index has been inverted such that higher scores reflect more transparent countries.
These laws may vary greatly; though they may not reflect great differences in government behavior. For instance, Iraq and Qatar receive quite similar Freedom House scores in the same year – both have similar legal protections for the freedom of speech – but Iraq receives a substantially lower HRV Index score.

Figure 10: HRV Index Plotted Against Freedom House Scores

A scatter plot of (standardized) Freedom House Freedom of the Press Scores and (standardized) HRV Index scores. Each observation corresponds to a country-year, and all country years with observations on both indexes between 1994 and 2004 are plotted. The dashed line runs through the origin at a 45 degree angle. Were the correlation between the two indexes perfect, all observations would lie on this line. Observation that lie above this line score relatively highly on the Freedom House Index (compared to HRV). Observations that lie below this line score relatively lowly on Freedom House. The dotted line is a lowess smoother. The marked observation is from Somalia in 1994.
A.2 The Relationship Between Newspaper Circulation and the HRV Index

Figure 11 depicts a scatter plot of the HRV Index against a commonly used measure of media market size – daily newspaper circulation per 1000 residents, as collected by the World Bank. Both measures have been standardized by subtracting the mean value from each observation and dividing by the standard deviation. The dashed line corresponds to a 45 degree angle; while the dotted line is a lowess smoother plotted over the data.

Figure 11 indicates that both indexes broadly agree on a set of low scoring countries. Countries with high scores on the HRV Index, however, vary greatly in their daily newspaper circulation numbers. The two indexes are correlated, though the disagreement over which countries are highly transparent is greater than between Freedom House and the HRV values.
A scatter plot of (standardized) World Bank Daily Newspaper Circulation per 1000 numbers and (standardized) HRV Index scores. Each observation corresponds to a country-year, and all country years with observations on both indexes between 1997 and 2004 are plotted. The dashed line runs through the origin at a 45 degree angle. Were the correlation between the two indexes perfect, all observations would lie on this line. Observation that lie above this line score relatively highly on the newspaper circulation index (compared to HRV). Observations that lie below this line score relatively lowly on newspaper circulation. The dotted line is a lowess smoother.
B Supporting Materials

B.1 Longitudinal Variation in HRV Scores

The economic and (to a lesser extent) political liberalization of the People’s Republic of China since the late-1970s and early-1980s through the present is well-known and widely commented upon.\textsuperscript{28} This liberalization has proceeded to such an extent that some authors refer to the PRC as exemplifying ‘authoritarian deliberation’ under which information is provided to the public, whose opinion is consulted in the policy-making process (He and Warren, 2011).\textsuperscript{29} One might therefore expect levels of transparency over the 1980-2008 period to have experienced a large improvement. Figure 12 demonstrates that – at least according to our measure – such an increase in transparency has indeed taken place. While there are few dramatic breaks in the time-series, the PRC’s HRV Index scores have improved steadily and significantly over time.

A dramatic decline in HRV Index scores is visible in Figure 13, which plots Somalia’s HRV Index scores over time. Throughout the 1970s and 1980s, Somalia was ruled by the nominally communist government of Mohamed Siad Barre.\textsuperscript{30} This government is associated with moderate to low HRV Index scores of between 0.5 and 0. By the late 1980s, this government faced increasing armed domestic opposition and infighting between clans once supportive of the government. Unrest grew such that, in 1991, Siad Barre was forced to flee Mogadishu in a tank, taking with him the foreign exchange reserves of the central bank (Prunier, 1996, 45). The collapse of the government ushered in a period of civil war during which Somalia became the paradigmatic example of a failed state. The collapse of

\textsuperscript{28}On the opening of China’s political system, see Landry (2008). On accountability in China see Tsai (2007). Kelliher (1997) notes that the Chinese discussion of local self-government largely focused on the replacement of corrupt officials with competent political outsiders.

\textsuperscript{29}China’s liberalization has also given rise to a greater openness to foreign influence. For a discussion of these changes, with specific reference to investment and climate change, see Lewis (2007, 163-165).

\textsuperscript{30}For an overview of the rise and fall of the Siad Barre regime, see Prunier (1996).
the Siad Barre government coincides perfectly with a dramatic – and statistically significant – fall in Somalia’s HRV Index score. In 1990, Somalia’s predicted HRV score fell from 0.046 to -0.111. The following year saw a still more dramatic drop to -0.686, followed by further declines. Somalia’s transparency score stabilizes around values of -1.66, the lowest in our sample.
Figure 13: Somalia HRV Scores Over Time

A longitudinal plot of Somalia’s HRV Index scores from 1980-2008. HRV Index scores are on the y-axis, while time is measured on the x-axis. Diamonds denote mean predicted index scores. Whiskers denote 95 percent highest posterior density intervals.