Integration:
An Empirical Assessment of Russia

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Abstract
In the process of implementing market reforms, many post-socialist countries have struggled to preserve economic and political integration. Using a statistical model of commodity trade, we quantify the evolution of economic integration observed among regions within Russia during 1995-1999, and then explore potential determinants of the patterns of integration we observe. Our measure of integration exhibits rich regional variation that, when aggregated to the national level, fluctuates substantially over time. In seeking to account for this behavior, we draw in part on theoretical models that emphasize the potential role of openness to international trade, regional disparities in income, and inflation volatility in threatening economic and political integration. Controlling for a host of additional regional- and national-level variables, we find a strong negative correspondence between openness to international trade and internal economic integration within Russia. We also find negative links but weaker links between integration and regional-income disparities and inflation volatility.

I. Introduction
A striking feature of the economic transitions undertaken by many post-socialist countries is the struggles they have encountered in preserving economic and political integration. Regarding economic integration, for example, following the economic reforms established by China in 1978, many regions within China erected inter-regional trade barriers and engaged in trade wars with each other (e.g., see Young, 2001, and references therein). Also, following the federal imposition of economic reforms in the early 1990s, many regional politicians within Russia violated federal laws mandating the free flow of goods across regions by taking such actions as the erection of border controls, the physical limitation of cross-border trade, and the issuance of local coupons designed to exclude non-residents from consuming locally produced goods (e.g., see Mitchneck, 1995; and Freinkman et al. 1999). Regarding political integration, economic transition has often coincided with political disintegration. For example, during the 1990s the former Czechoslovakia split into two independent countries, the former Yugoslavia split into six countries, and the Former Soviet Union (FSU) split into fifteen independent countries. Furthermore, regions in many of the newly independent countries within the FSU such as Azerbaizhan, Georgia, Moldova, Russia and the Ukraine have subsequently taken measures that have undermined the political integration of their countries. For example, starting in the early 1990s, many ethnic regions within Russia engaged in separatist activities that included declarations of sovereignty, refusals to send conscripts to the federal army, assertions of the right to print their own currencies, and assertions of control over natural resources that had been under federal jurisdiction. This in turn has encouraged many non-ethnic regions to engage in similar acts of separatism (see Treisman 1997, 1999a). Perhaps most notably, many Russian
regions have chosen to delay or even withhold mandatory tax payments owed to the federal government. In particular, the massive decline in federal tax receipts between 1992 and 1998 resulted in large part from the growth in late payments and non-payments from regional budgets (OECD 1997 and 2000; and Treisman 1999b).

In order to better understand forces that may be responsible for undermining economic and political integration in post-socialist nations, we use a rich regional data that enables us to quantify and attempt to account for patterns of regional economic integration observed within Russia. We quantify economic integration between 1995-1999 using a statistical model of commodity trade developed by Engel and Rogers (1996) and Parsley and Wei (1996). The model is designed to evaluate whether the behavior of commodity prices in a particular region vis-à-vis the region’s potential trading partners is consistent with the behavior one would expect to observe if the region in question was actually open to regional trade. We use the model to produce a yes/no signal regarding openness to domestic trade for 72 of Russia’s now 89 regions (complete data on the remaining regions are unavailable). We produce such a signal for each region on a monthly basis. Summing signals over regions yields an aggregate measure of the proportion of regions deemed to be integrated at a given point in time; the aggregate measure we obtained is illustrated in Figure 1.

There is striking temporal variation in the aggregate characterization of integration provided in Figure 1. Beginning in 1995, an initial period of widespread integration gradually gave way to a period of disconnectedness in 1996 and 1997, which seems to have subsided by 1998. At the disaggregated level, there is also striking variation in patterns of integration observed across regions. For example, the Sakha Republic was isolated from its internal trading partners throughout the sample period; Krasnoyarsk and Udmurtia maintained strong internal connections throughout the period; Ivanovo was integrated initially but subsequently became isolated; and Arkhangelsk was isolated initially and then gradually became integrated.

In seeking to account for these patterns of economic integration, we rely in part on guidance from theoretical characterizations of economic and political integration. In particular, Alesina et al. (2000) and Casella (2001) present models in which trade liberalization can pose a threat to political integration by making regions within a political union less reliant on fellow members as trading partners, and also as sources of financing for public goods and institutions. In the case of Russia, trade liberalization potentially poses a particularly important threat because pitfalls associated with inter-regional trading, such as late payments and high and arbitrary federal and regional taxes, are particularly severe (Ericson, 2001). In addition, Bolton and Roland (1997) caution that disparities in regional incomes can weaken internal political ties in regions that are sufficiently poor and sufficiently rich. If a region is sufficiently rich, it has the fiscal capacity to finance its own public goods independently, and at some point may lose
its willingness to subsidize the provision of public goods in the rest of the country. In contrast, if a region
is sufficiently poor, it may break off from the federation if it becomes dissatisfied with the extent of
federal redistribution it receives. Finally, models developed by Lucas (1972), Barro (1972), and
Sheshinski and Weiss (1977, 1983) all highlight how inflation volatility can weaken internal economic
integration by making the comparison of relative prices across regions more difficult.

In many respects, Russia offers a particularly useful laboratory for evaluating the influence of
these potential sources of disintegration. During its transition, it has made great strides in opening its
economy to world markets; it has experienced dramatic and persistent disparities in standards of living
across regions; and it has weathered severe inflationary spikes. Our purpose is to investigate empirical
links between these conditions and the patterns of economic integration noted above.

We recognize that several of the theories noted above specifically address the issue of political
integration, while our focus is on economic integration. Nevertheless, we believe these theories are of
potential relevance in this context, due to natural links between economic and political integration. For
example, external trade coalitions arising from trade-liberalization reforms do not merely afford regions
with the opportunity to self-finance regional public-good provision; they also provide regions with the
option of pursuing trade relationships with regions outside their own country, potentially lessening the
importance of internal trading partners. Thus if trade liberalization was in fact actively working to
produce political disintegration within a country, we might well expect to observe economic
disintegration (in the guise of a decreased importance of domestic trading partners) in parallel, or even as
a symptom of political disintegration that becomes apparent well in advance of an actual breakup. In the
case of Russia, we find empirical support for this link: specifically, we find a strong negative association
between our measure of economic integration and Treisman’s (1999b) measure of regional tax
withholdings to the federal government, which serves as a useful measure of political disintegration (see
Section II for details).

We use a reduced-form statistical framework to characterize economic integration at the regional
level. This involves modeling regions as individual entities that base decisions regarding participation in
trade with domestic partners in a given time period on a collection of region-specific and national-level
variables (described in detail in Section III). National-level variables consist of annual measures of
openness to international trade and inflation volatility, as well as variables that quantify strike activity and
the pervasiveness of poverty at the aggregate level. Region-specific variables are of two types: time-
varying and fixed. The time-varying measures control for regional patterns of poverty and fluctuations in
transport costs (which are included to quantify a potential detriment to domestic trade). The fixed
variables include two measures of income disparities (one that measures the disparity between regional
and national median incomes for rich regions, and one for poor regions). They also include regional measures of initial income, education levels, and language diversity.

Our analysis indicates the presence of a strong negative link between openness to international trade and economic integration within Russia. Under our baseline estimates, a one-standard-deviation increase in our aggregate measure of openness to trade is associated with a 6-7% decline in the probability that a region will be linked with its internal trading partners in any given period. Results regarding regional income disparities are mixed. The Bolton-Roland model suggests that a region will tend to have a lower propensity towards integration as the absolute difference between its median income and the national median income increases. We find that this pattern holds only for regions with relatively low median incomes. However, we do find strong negative and significant links between integration and initial income levels and initial educational levels; thus there is some evidence of the negative association between regional prosperity and integration one would expect in light of the Bolton-Roland model. Finally, we also find a negative link between inflation volatility and integration, but the link is of marginal statistical significance.

A substantial literature has arisen out of attempts to measure and explain economic integration in post-socialist economies. Gardner and Brooks (1994) and De Masi and Koen (1996) studied this problem for the early phase of Russia’s reform and found that large differences in inter-regional prices could not be explained by distance. Since arbitrage opportunities for tradable goods do not persist in inter-regional markets within developed countries, both studies concluded that Russia’s market integration was limited. In Berkowitz and DeJong (1999) we found that a reason for this high inter-regional price dispersion was the behavior of pro-Communist and anti-market regions. In Berkowitz and DeJong (2001), we conducted an aggregate-level study that sought to account for the dynamic behavior of economic integration in Russia; we found that aggregate measures of economic integration exhibit a strong negative relationship with openness to international trade, and have no significant correspondence with aggregate inflation volatility. Young (2001) documented and explained economic disintegration in China following its 1978 reforms, and noted that it too coincided with an increased openness to world markets. Finally, in a detailed econometric analysis of Poland’s convergence to economic integration following its price liberalization in the early 1990s, Konienczny and Skrzypacz (2000) found that inflation and inflation volatility had a substantial impact on the speed of convergence to economic integration.

II. Measuring Integration

If regional markets are integrated within a country, arbitrageurs and traders can move goods from regions in which prices are low to regions in which they are high so long as transport costs are not prohibitive. Thus if transport costs are increasing in distance, so too will be regional price differentials.
More formally, let $Q_{ij}(t)$ denote the percentage price differential for some tradable good sold in regions $i$ and $j$ at some date $t$: $Q_{ij}(t) = \text{abs}(\log(P_i(t)/P_j(t)))$. Also, let $d_{ij}$ represent the distance separating regions $i$ and $j$, where inter-regional transport costs are increasing in $d_{ij}$. Lacking barriers to inter-regional trade, arbitrage opportunities exist at some date $t$ when $Q_{ij}(t)$ is greater than or equal to transport costs. When $Q_{ij}(t)$ is less than transport costs, moving the good between these two regions is not profitable, and inter-regional trade does not occur.

To quantify the relationship between distance and transport costs, we follow Engel and Rogers (1996) in employing Krugman’s (1991) transport cost model, in which $1 - 1/(1+ d_{ij})$ is the share of a good that depreciates when it is moved between regions $i$ and $j$. Arbitrageurs can profitably buy goods in region $j$ and resell these goods in region $i$ when $P_i/(1 + d_{ij}) \geq P_j$; the same arbitrageurs can profitably buy goods in region $i$ and resell in region $j$ when $P_j/(1 + d_{ij}) \geq P_i$. Given an integrated internal market, there will be trade between regions $i$ and $j$ when the relative price $(P_i/P_j)$ fluctuates outside the band $[1/(1+ d_{ij}), (1+ d_{ij})]$. This band is increasing in $d_{ij}$; thus a testable implication of this model is that the variance of $Q_{ij}$ is increasing in inter-regional distance when internal markets within a country are integrated.

We measure movements in inter-regional price dispersion using a monthly price index that measures the cost of a weighted basket of basic food goods. The index spans January 1995 through December 1999, and was compiled by the Russian statistical agency Goskomstat (unpublished). The index includes observations from 72 Russian cities including Moscow and St. Petersburg, and capital cities in 70 additional regions. (There are 89 regions in Russia; the data we lack are from war-torn regions in the Caucasus, and from many of the autonomous okrugs and krais).

Let $t = 1, 2, \ldots, M$ denote a particular month and year in which the price index is computed, and let $\sigma_{ij}(s)$ denote the standard deviation of $Q_{ij}(t)$ calculated over the twelve-month sub-period indexed by $s$. To measure movements in price dispersion, we calculate $\sigma_{ij}(s)$ for every possible $(i,j)$ combination such that $i \neq j$, and for every possible twelve-month sub-period in the time period spanned by our data (we also considered shorter and longer sub-periods and obtained similar results). There are 60 months spanned by our price index, so there are 49 possible twelve-month sub-periods, including five that begin in January. The sub-periods beginning in January of course contain all of the twelve months contained in a particular year; thus it is these that we use in Section III to quantify integration on an annual basis.

To measure market integration for region $i$ in sub-period $s$, we use OLS to estimate

\[
\sigma_{ij}(s) = \alpha_{ij}(s) + \beta_{ij}(s)\log(d_{ij}) + u_{ij}(s), \quad i \neq j,
\]

where $\alpha_{ij}(s)$ is the estimated intercept, $\beta_{ij}(s)$ is the coefficient for log distance from region $i$ to all other regions, and $u_{ij}(s)$ is an error term. Since inter-regional price volatility in integrated economies should be increasing in log distance, we deem region $i$ to be integrated during sub-period $s$ when $\beta_{ij}(s)$ is estimated.
as positive and statistically significant.\textsuperscript{1} We define significance at the 10% level in the results reported below; the patterns identified in this manner are robust to alternative choices. Standard errors used to evaluate significance are heteroscedasticity consistent (White, 1980).\textsuperscript{2}

Figure 1 plots the percentage of regions deemed to be integrated in each sub-period contained in the period January 1995-December 1999. The initial plotted value indicates that roughly 79\% of the regions in our sample were deemed to be integrated over the period January 1995 – December 1995; the second value spans the period February 1995 – January 1996; etc. The percentage of integrated regions fell from roughly 79\% to 57\% between January 1995 and June 1996; it then climbed steeply until January 1997, where it attained almost 89\%, fell to roughly 77\% in June 1997, and then gradually rose to 93\% in January 1998. Integration almost always exceeds 90\% through the end of the sample.

While not apparent in Figure 1, there are substantial differences in the particular patterns of integration followed by Russia’s regions. Among the 72 regions, 55 are integrated in at least four of the five sub-periods beginning in January, and three are not integrated in four of the five sub-periods. Eight regions exhibit an “out-in” trajectory (isolated in at least the first two January sub-periods, and integrated in at least two of the remaining three). Two regions exhibit an “in-out” trajectory (integrated in at least the first two January sub-periods, and isolated in at least two of the remaining three). The remaining four regions exhibit oscillating patterns of integration.

These results portray an optimistic picture of Russia: notably, the internal market has been highly integrated since 1998. Nevertheless, aspects of this integration trajectory are puzzling at first glance. During 1996-97 the federal government had successfully stabilized the ruble, and by 1997 the sustained decline in output Russia suffered during transition had ceased. Nevertheless, market integration clearly deteriorated over this period. In contrast, real income fell and inflation spiked following the financial crisis in the summer of 1998, yet integration improved throughout 1998, and this improvement was sustained through 1999.

Before describing the details of our attempt to explain these regional integration dynamics, it is important to note that we could use several alternative methods to quantify integration. For example, drawing upon the literature devoted to evaluating relationships between inflation and relative price

\textsuperscript{1} A potential problem with this method of measuring integration is that a region may fail to appear integrated not because it chose to be this way, but because a sufficiently large proportion of its neighbors did. To explore this possibility we considered an alternative procedure that involves two steps. The first step is as explained above. In the second step, we re-estimated (1) for each region i by measuring its price volatility vis-à-vis only those regions deemed to be integrated in the first step. Use of this alternative procedure had a minor impact on the shares reported here.

\textsuperscript{2} Potential concerns regarding heteroscedasticity also prompted us to estimate (1) using logged values of $\sigma_{ij}(s)$. The correlation between the measure of integration obtained using logs and that presented in Figure 1 is 0.96, and the results presented in Section IV are robust to the use of the measure based on logs.
variability, we could use “coefficient of variation” statistics, which provide an aggregate characterization of price dispersion across regions at a given point in time (recent additions to this literature include Silver and Ioannidis, 2001; and Konieczny and Skrzypacz, 2000). As it turns out, the correlation over time between the aggregated integration measure illustrated in Figure 1 and the corresponding coefficient of variation measure (calculated on a month-by-month basis) is −0.713, so these measures exhibit close correspondence. However, while coefficient-of-variation statistics only provide an aggregate characterization of price dispersion, at the disaggregated level our measure provides a measure of integration for each region at any given point in time. This feature will enable us to bring aggregate and regional variables to bear in seeking to characterize internal integration across regions and over time.

Alternatively, we could characterize integration using inter-regional trade-share data. Unfortunately, such data are available within Russia for a small set of goods measured on an annual basis between 1995 and 1998 (meat including butter, cheese, meat including chicken, sausage, sugar, and vegetable oil; source: Goskomstat, 2000). To evaluate the correspondence between our integration measure and these data, we fit our 0-1 measures calculated for each region over the periods (January 1995 – December 1995); (January 1996 – December 1996); … (January 1998 – December 1998) on these regional trade-share data using a probit model. Controlling for year-specific effects, we found trade shares to be quantitatively and statistically (at the 2% level) significant in this analysis; specifically, a one-standard deviation increase in the share of inter-regional trade corresponds with an 18.9% increase in the probability of an affirmative indication of integration. In this case, our regional integration measure is advantageous since it is based on a relatively broad set of food goods (19 during 1995-96 and then 25 starting in 1997), and it is measured on a monthly basis.

Finally, it would also be attractive to work with regional measures of political integration, e.g., as in Treisman (1999b), who interprets growing regional-government tax arrears as indicating the unwillingness of regional governments to cooperate with the federal government in financing public goods and services, and thus as providing a measure of the breakdown of political integration. Treisman’s data quantifies tax arrears at the regional level in November of 1996 and 1998. Normalizing arrears by the share of taxes paid in the previous year, we fit our integration measures calculated for (January 1995 – December 1995) and (January 1996 – December 1996) on 1996 arrears; and our measures calculated for (January 1997 – December 1997) and (January 1998 – December 1998) using a probit model, and once again found close correspondence. Controlling for year-specific effects, the coefficient we obtained was significant at the 6% level, and implies that a one-standard-deviation increase in arrears corresponds with a 14.9% decrease in the probability of an affirmative indication of integration.
In sum, our integration measure corresponds closely with alternative measures of both economic and political integration, while retaining various advantages. In the next section, we describe our attempt to account for its behavior.

III. Accounting for Integration

Here we discuss the explanatory variables and statistical framework used to characterize integration. As noted in the introduction, the theoretical frameworks of Alesina et al. (2000) and Casella (2001) predict that increased openness to international trade can threaten political integration by reducing the dependence of domestic regions on domestic trading partners and the federal government for the provision of public goods, institutions, and trade relations. Moreover, Ericson’s analysis (2001) suggests that this may be a particularly important concern in Russia, due to the presence of pervasive distortions associated with internal trade. Finally, trade liberalization has been an important component of the economic reforms Russia has adopted in transition, thus we have an ideal opportunity to assess the impact on integration of a meaningful policy change along this dimension.

The theoretical model developed by Bolton and Roland (1997) predicts that the emergence of large inter-regional income disparities can also threaten political integration by increasing the demand for fiscal redistribution from the poorest regions, while placing an unacceptably large redistribution burden on the richest regions. Such disparities have been striking in Russia. For example, between 1993 and 1997, real income grew at a staggering annual average rate of 15.7% in Moscow, while it shrank at annual average rate of -9.0% in Bryansk (Berkowitz and DeJong, 2002). Regional variation in poverty is also staggering: more than 75% of the populations in Chita, Kalmykia and Tuva lived under the poverty line in 1999, while less than 25% of the populations in Moscow and Samara were impoverished (Goskomstat, 2000b). Finally, there is substantial fiscal redistribution across regions within Russia, and the relatively rich regions are expected to bear the financial burden of this activity (Treisman, 1999a; Freinkman et al., 1999).

There is a large empirical literature linking regional price dispersion with aggregate inflation and inflation volatility (e.g., Lach and Tsiddon, 1992; Parsley, 1996; DeBelle and Lamont, 1997; Van Hoomissen, 1998; and Konieczny and Skrzypacz, 2000), thus we too consider the impact of inflation volatility on integration. Once again, Russia provides a useful laboratory for doing so. Following the price liberalization in 1992, consumer prices increased over 20-fold on an annual basis during 1992 and 1993 (Gardner and Brooks, 1994). A period of successful price stabilization then ensued: from 1994 through 1997, consumer price inflation fell steadily from an annual rate of 115% to 10%, and monthly fluctuations in the inflation rate were also reduced dramatically. However, following the financial crisis in August of 1998, inflation in September of 1998 spiked to an annualized rate of 390%, remained
volatile through the remainder of the year, and then fell both in terms of level and volatility in 1999, when it averaged 31% (SITE, 2001).

The general finding in the inflation-volatility literature is that price dispersion is positively related with inflation volatility. One explanation for this comes from Lucas (1992), who models shopkeepers as making relatively large price increases when they perceive price shocks as signaling relatively high ratios of national to regional inflation rates. Because shopkeepers’ pricing strategies are also related to other factors including own costs, market costs, market demand, and inflation persistence, and since these other factors differ across regions, movements in aggregate inflation rates can generate increased regional price dispersion. (Menu costs are an alternative source of such relationships; e.g., see Barro 1972; and Sheshinski and Weiss 1977, 1983). To the extent that aggregate inflation volatility makes the comparison of relative prices across regions more difficult, it may be that increases in aggregate inflation volatility tend to weaken internal economic integration as well.

We measure openness to international trade at the national level on an annual basis using data on the volume of trade conducted outside the Commonwealth of Independent States, measured as a percentage of GDP (source: Stockholm Institute for Transition Economics – SITE, 2001). Regarding disparities in regional incomes, the variable featured in the model of Bolton and Roland (1997) is the disparity in incomes between the median voter in each region and the national median voter. In their construct, the median voter is the person with the median income; in reality, this will be the case only if voting preferences are determined solely by the income differences their model emphasizes. Nevertheless, using differences between regional and national median-income levels as a proxy for differences in the incomes of median voters in this case seems reasonable. Ideally, we would measure disparities in median incomes on an annual basis; unfortunately, information on regional income distributions is available only for 1995 (source: Goskomstat, 1996). Thus income disparities are measured as regional fixed effects. We have two disparity measures: one for poor regions, and another for rich regions. To construct the measure for poor regions, we assign the value of zero to all regions whose median income lies above the national median income (the “rich” regions). For the remaining regions, the measure we use is the logged absolute difference between the regional and national median incomes, measured as a percentage of the national median income. The measure for rich regions is constructed analogously. We distinguish between rich and poor regions since the two types of disparities have potentially differing relationships with integration. We measure inflation volatility using monthly

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3 We thank Anthony Shorrocks for informing us about data sources on income distribution, and for making these data available to us.

4 We also experimented with alternative measures in which “rich” (“poor”) regions had to have median incomes that exceeded (lied below) the national median income by certain thresholds (10%, 20%, etc.) before qualifying as being rich (poor). These experiments yielded results similar to those reported here.
CPI data (source: SITE 2001). Inflation volatility is calculated as the standard deviation of the monthly inflation rate observed during the year (for a note regarding the use of inflation as an additional explanatory variable, see footnote 7).

In addition to these variables, we control for the potential influence of other national and regional economic variables. Somewhat related to measures of regional income disparities are the annual national and regional percentages of the population whose incomes lie below the official poverty line (source: Goskomstat, 200a,b). We also control for the annual inflation rate of freight transport costs observed for each year in each region, since relative surges in transport costs may be an important explanatory variable in accounting for internal integration (source: Goskomstat, 200a,b). We use annual regional strike data in an attempt to control for the possible influence of regional public discontent. This is measured as the workdays lost in strikes per thousand workers (source: Goskomstat 2000a,b). Finally, we control for differences in regional population characteristics using initial education levels, initial regional language diversities, and initial regional incomes.\(^5\) We measure initial education using the share of the population fifteen years old and higher as of 1994 that had completed high school and received some post-secondary education. We measure initial regional language diversity by computing the percentage of people in 1994 that do not consider Russian to be their native language. The education and language statistics are both taken from the 1994 Goskomstat micro-survey of regions (Goskomstat, 1995a,b). We measure initial regional income by calculating the ratio of money income per capita to the cost of a uniform basket of 25 basic foods observed during 1993:IV (source: unpublished Goskomstat data). We choose 1993:IV because this is the earliest quarter for which we have been able to gather comprehensive data for food baskets and per capita income data. The food basket is a useful deflator because more than half of Russian household expenditures in the 1990s were allocated to food goods (Goskomstat 2000c, p.167).

The mechanical structure of our empirical model is as follows. Since our time-varying explanatory variables are generally available only on a yearly basis, our dependent variable – regional market integration – is measured annually as well. This is done by evaluating the statistical significance of the distance coefficient \(\beta_i(s)\) calculated for each region over the five twelve-month sub-periods January-December included in our data set. If \(\beta_i(s)\) is estimated as statistically significant at the 10% level over the 1995 January-December sub-period, we assign the value of one to the 1995 entry for region \(i\); otherwise we assign the value of zero.\(^6\) So our integration measure consists of 72 zeros or ones for

\(^5\) We also used data on defense employment and initial production potential as additional regional controls. However, these variables had neither significant statistical nor quantitative relationships with integration. Moreover, we lacked complete regional coverage of these variables, thus they were not incorporated in the analysis presented here.

\(^6\) As mentioned above, we also experimented with alternative significance levels (e.g., 5%, 15% and 20% levels) in defining market integration, and obtained results that closely mirror those presented below.
Hereafter, we denote our integration measure by the 360x1 vector \( y \), the measure calculated for sub-period \( s \) by the 72x1 vector \( y(s) \), and region \( i \)’s entry for sub-period \( s \) by the single value \( y_i(s) \).

Regarding the explanatory variables, we have five annual observations on the four national-level variables (openness to international trade, inflation volatility, national poverty and strikes); five region-specific variables that are fixed over time (the rich and poor income-disparity variables, initial income, initial education; and initial language diversity), and two region-specific variables that vary over time (regional poverty and transport-cost inflation). Let the national-level variables be contained in the 360x4 matrix \( X_N \), which consists of five 72x4 sub-matrices \( X_N(s), s = 1995, 1996, \ldots, 1999 \). Each column of \( X_N(s) \) contains a set of 72 identical observations on a national-level variable observed in year \( s \). Let the fixed region-specific variables be contained in the 360x5 matrix \( X_{RF} \), which consists of five identical 72x5 sub-matrices \( X_{RF}(s) \) (i.e., \( X_{RF}(1995) = X_{RF}(1996) = \ldots = X_{RF}(1999) \)). Row \( i \) of \( X_{RF}(s) \) contains the income-disparity variables, initial income, education and language diversity observed for region \( i \). Let the region-specific variables that vary over time be contained in the 360x2 matrix \( X_{RV} \), which consists of five 72x2 sub-matrices \( X_{RV}(s) \), whose \( i \)th row contains period-\( s \) observations on region \( i \)’s poverty level, and transport costs. Finally, let \( X \) denote the entire collection of explanatory variables obtained by vertically concatenating \( X_N, X_{RF}, X_{RV} \) and a vector of ones, and \( X_i(s) \) denote the row vector containing the explanatory variables observed in region \( i \) in year \( s \).

We model the probability that region \( i \) is deemed to be integrated in year \( s \) – \( \text{prob}(y_i(s) = 1) \) – using a standard probit model:

\[
\text{prob}(y_i(s) = 1) = \int_{-\infty}^{X_i(s)B} \phi(t)dt = \Phi(X_i(s)B),
\]

where \( \Phi(.) \) denotes the c.d.f. of the standard normal distribution (use of a logit model yields similar results). The parameter vector \( B \) is estimated via maximum likelihood using GAUSS.

Before discussing model estimates, some notes regarding the issue of simultaneity are in order. In our view, the explanatory variables fall into three classes: those for which simultaneity is clearly an

\[\text{We note that since our data span five time periods, we are constrained to consider at most four national-level variables (in addition to the constant); including additional variables (e.g., time dummies, inflation, etc.) yields a rank reduction in } X, \text{ rendering } X'X \text{ non-invertible. It is for this reason that we do not consider both inflation and inflation volatility as explanatory variables in our analysis. (We chose to drop inflation in favor of the remaining variables in the analysis presented below because its a priori relationship with integration is arguably weakest. Moreover, its empirical relationship is also weak: replacing any of the four national-level variables considered below with inflation yields an insignificant parameter estimate for the coefficient on inflation.) In contrast, since our data include 72 regions, we do have further ability to control for fixed region-specific effects. Controlling for such effects by augmenting } X \text{ with regional dummies has no substantive influence on the results presented below.}\]
issue; those for which simultaneity is unclear *a priori*; and those that are exogenous. The variables we take to be clearly exogenous are inflation volatility; regional transport-cost inflation; initial income in 1993:IV, initial education and initial language diversity (since they all pre-date our sample), and the 1995 measures of regional income disparities (since it is measured at the beginning of our sample). The variables with the clearest potential for simultaneity include the national and regional poverty measures; and the national strikes. For these variables, we use one-year lagged values as instruments. Finally, simultaneity is less clear for the aggregate international trade. For example, under the theoretical constructs of Alesina et al (2000) and Casella (2001), participation in internal markets responds negatively to exogenous changes in openness to international trade. Alternatively, it could be that participation rates in internal and international markets respond simultaneously to exogenous changes in other variables. Given our uncertainty, we initially estimated B using contemporaneous values of the aggregate international-trade measure, and then re-estimated B using lagged values. The two sets of results we obtained are closely comparable. Given this, we report the full set of results obtained using the contemporaneous international-trade measure in Table 1.

**IV. Empirical Results**

Two sets of results are reported in Table 1. The first set was obtained using each of the explanatory variables, including contemporaneous values of the international-trade measure. The second set was obtained by excluding several of the regional control variables that turned out to be insignificant both statistically and quantitatively: language diversity, transport-cost inflation, poverty, and strikes. (Exclusion of these variables yields a likelihood ratio statistic of 2.33, which has a p value of 0.67; it also yields minor changes in the remaining estimates.) The last column of the table reports the quantitative significance of our estimates, which is assessed by calculating the impact on prob(γ_i(s) = 1) of a one-standard-deviation increase in the corresponding variable.

Among the control variables, the national measure of poverty is most notable: its estimated coefficient is negative and significant at the 1% level in both sets of estimates, and its measure of quantitative significance is approximately 10 percentage points. The regional measures of educational attainment and initial incomes also have negative and statistically significant coefficients, with corresponding measures of quantitative significance of approximately 5.5 percentage points. And as noted, the remaining control variables yield negligible explanatory power, and are clearly excludable.

Turning to the theoretically motivated variables, the coefficient associated with openness to international trade is negative and significant at the 1% level in both models, and its quantitative significance is in the 6 – 7 percentage point range. (Working with the lagged measure of trade also yields a coefficient estimate that is negative and significant at the 1% level, with a corresponding quantitative
significance measure of 15 percentage points.) These results are consistent with the theoretical predictions of Alesina et al. (2000) and Casella (2001).

Regarding disparities in regional median incomes, the Bolton-Roland model suggests the existence of a negative relationship between regional integration and the absolute difference in incomes between regional and federal median voters. This relationship is evident only in regions with low median incomes: quantitative significance measures associated with the low-median-income variable are approximately 5 percentage points, and the coefficient is statistically significant at the 10% level in the restricted model. Coefficients associated with the high-median-income variable actually turn out positive, although their p values exceed 20% in both models. So the results along this dimension are mixed. However, the clear negative links between integration and the regional measures of educational attainment and initial incomes indicate the presence of weaker internal ties among relatively prosperous regions, as one would expect in light of the Bolton-Roland model.

Finally, we observe the expected negative relationship between integration and inflation volatility. However, the strength of the relationship is marginal: p values are in the 15% – 20% range, with associated measures of quantitative significance of approximately 4 percentage points.

In considering these results, it is important to be mindful of the fact that we do not observe economic integration directly. Rather, the results are based on the measure of integration generated by the statistical model of commodity trade outlined in Section II. This raises the possibility that the results are tainted by inconsistency arising from potential mis-measurement. In an effort to quantify the extent of potential mis-measurement, and to evaluate the potential influence of mis-measurement on these results, we obtained a second set of estimates using the mis-measurement model developed by Hausman, Abrevaya, and Scott-Morton (1998). The model is a modification of the probit model (2) that incorporates two additional components. Recall that \( y_i(s) \) denotes our measure of whether region \( i \) is integrated in year \( s \). Letting \( y^*_i(s) \) denote the corresponding actual integration status of region \( i \) in year \( s \), the additional components are given by

\[
(3) \quad \alpha_0 = \Pr(y_i(s) = 1 | y^*_i(s) = 0), \\
(4) \quad \alpha_1 = \Pr(y_i(s) = 0 | y^*_i(s) = 1).
\]

Thus \( \alpha_0 \) denotes the probability that region \( i \) is spuriously classified as being integrated in period \( s \), and \( \alpha_1 \) denotes the probability that the region is spuriously classified as not being integrated. Taking into account these probabilities of mis-measurement, (2) becomes

\[
(5) \quad \text{prob}(y_i(s) = 1) = \alpha_o + (1 - \alpha_o - \alpha_1) \Phi(X_i(s)B).
\]
We estimated this model using non-linear least squares, which involves minimizing

\[
(6) \quad \sum (y_i(s) - \text{prob}(y_i(s) = 1))^2
\]

over \((\alpha_0, \alpha_1, B)\); estimates are presented in Table 2. We were unable to obtain estimates of this model for the full set of variables considered above, thus Table 2 presents results only for the restricted set of variables used to generate the second set of results presented in Table 1. Given the minimal explanatory power of the excluded variables in the probit model, their exclusion in this mis-measurement model seems warranted (indeed, their lack of explanatory power seems to be the source of the difficulties in estimating the fully-specified model).

Comparing the results presented in Tables 1 and 2, problems associated with mis-measurement seem minor in this application. First, note that the estimated value of \(\alpha_0\) (0.003, or 0.3\%) is small and statistically insignificant. In contrast, the estimated value of \(\alpha_1\) is non-trivial (6.9\%) and statistically significant at the 5\% level, indicating that our integration measure tends to underestimate the true extent of economic integration in Russia. However, this carries only minor implications for the additional coefficient estimates. In particular, the estimated signs of each coefficient are the same in both models. Also, the pattern of statistical significance across variables is the same in both tables, with one exception: the coefficient on high median income was insignificant in the probit model, but is significant at the 10\% level in the mis-specification model. Finally, quantitative significance measures are somewhat lower in the mis-specification model (again, with high median income serving as the exception), but remain substantial. For example, a one-standard-deviation increase in the trade-openness variable corresponds with a 5.5 percentage point decrease in the estimated probability of being integrated in this case, as compared with a 7.2 percentage point decrease in the probit model. In sum, although we are only able to measure regional patterns of economic integration rather than observe these patterns directly, associated problems arising from potential mis-measurement do not seem to be a serious concern in this application.

We conclude with a note regarding the 0-1 measure we use to characterize regional integration. Recall how this measure is constructed: region \(i\) is deemed to be integrated in period \(s\) (and thus \(y_i(s)\) is set to 1) if its coefficient on logged distance is estimated as positive and statistically significant in regression model (1); otherwise \(y_i(s)\) is set to 0. Among the attractive features of this approach is the straightforward means in which the measure may be aggregated over regions to achieve a dynamic characterization of the extent of integration at the national level, e.g., as is done in Figure 1. However, the conversion of a continuous measure of the importance of distance in accounting for regional patterns
of price dispersion into a 0-1 measure imposes a certain loss of informational content. Our own view is that this imposition is appropriate: region A should not necessarily be deemed to be twice as strongly integrated with its domestic trading partners than region B just because A’s standardized distance coefficient is twice as large as B’s. Nevertheless, the robustness of the inferences based on our 0-1 measure is an important consideration, thus an exploration of the impact of this imposition seems worthwhile.

To do this, we respecified \( y(s) \) as the standardized distance coefficient (i.e., the t statistic of the distance coefficient) estimated in (1). We then estimated the relationship between this alternative measure of integration and our explanatory variables using OLS; results are given in Table 3. Several features of these results mirror those obtained using the 0-1 measure. Notably, the negative and statistically significant relationship observed between integration and openness to international trade is evident once again; the same is true for the regional measures of educational attainment and initial incomes, and the national measure of poverty. Also evident once again is the weak statistical link between integration and inflation volatility; indeed, the link is much weaker in this case.

The biggest difference in results involves the low-median-income measure, whose coefficient is positive and significant in this case. This is perhaps due to a second difference in results, this time involving the regional poverty measure. Previously, the coefficient on this variable was essentially zero; in this case it is negative and significant. Thus the results based on the 0-1 measure suggest that sub-par median regional incomes pose the greatest relative threat to regional integration, while the results in this case suggest that the incidence of regional poverty poses the greatest relative threat. In either case, there is an evident link in the data between regional economic stagnation and a weakening of economic integration. Again, this is the relationship one would expect in light of the Bolton-Roland model.

As a final comparison, statistical links between the explanatory variables and this alternative measure of integration are generally stronger than was the case with the 0-1 measure, with transport-cost inflation and inflation volatility providing the exceptions. Thus there does appear to be some deterioration in informational content associated with the use of the 0-1 measure. However, the deterioration is relatively minor, as are differences in the general characterizations of integration that emerge from the use of either measure.

V. Conclusions

The purpose of this paper has been to use the case of Russia in order to better understand forces that may be responsible for undermining economic and political integration in post-socialist countries. In order to quantify integration dynamics we have used a statistical model of commodity trade. To account for these dynamics we have relied on guidance from theoretical work that highlights the potentially
destabilizing impact on economic and political integration of openness to international trade, regional income disparities and inflation volatility.

Our analysis has produced four main findings. First, at the aggregate level, Russia has experienced substantial fluctuations in the share of internally integrated regions. Specifically, following an initial period of widespread integration in 1995, Russia gradually became fragmented in 1996 and 1997, and then became re-integrated by 1998. Underlying this aggregate behavior is a considerable degree of diversity in the specific integration patterns observed across regions.

Second, openness to international trade seems to have worked to undermine Russia’s internal economic coherence. This result provides an important caveat to the standard policy prescription that trade liberalization is critical because it forces enterprises to restructure and allows regions to reap gains from trade. Our results suggest that regions within Russia have actively substituted between domestic and international trading partners in the latter half of the 1900s. This result complements Young’s (2001) observation that China’s internal market became more fragmented during a period when China expanded its trade on world markets.

Third, striking disparities in regional incomes also seem to have undermined integration within Russia. This is most clearly evident among relatively poor regions, although it is also the case that regions with relatively high initial incomes on average exhibit lower propensities towards internal integration. This seemingly presents federal policymakers interested in preserving national unity with a difficult policy challenge: address the plight of relatively poor regions without imposing a politically unacceptable burden in doing so on relatively rich regions.

Finally, the period of high inflation volatility associated with the financial crisis does seem to have had an impact on internal integration within Russia, but substantially less than what one might expect given the severity of the episode.
References


______________ *Raspredelenie naselenia Rossii po vladeniiu iazykami ( po dannym mikroperepisi naseleniya 1994 g.).* Goskomstat Rossi, Moscow, 1995b.


Table 1 – Probit Estimates

<table>
<thead>
<tr>
<th></th>
<th>Coefficient</th>
<th>Standard Error</th>
<th>t statistic</th>
<th>p value</th>
<th>Quantitative Significance</th>
</tr>
</thead>
<tbody>
<tr>
<td>Constant</td>
<td>12.515</td>
<td>4.415</td>
<td>2.830</td>
<td>0.005</td>
<td>0.0%</td>
</tr>
<tr>
<td>Education</td>
<td>-0.064</td>
<td>0.024</td>
<td>-2.620</td>
<td>0.009</td>
<td>-5.4%</td>
</tr>
<tr>
<td>Initial Income</td>
<td>-0.102</td>
<td>0.040</td>
<td>-2.580</td>
<td>0.010</td>
<td>-6.2%</td>
</tr>
<tr>
<td>Language</td>
<td>0.007</td>
<td>0.006</td>
<td>1.040</td>
<td>0.298</td>
<td>2.4%</td>
</tr>
<tr>
<td>High Median Income</td>
<td>0.164</td>
<td>0.130</td>
<td>1.250</td>
<td>0.210</td>
<td>4.5%</td>
</tr>
<tr>
<td>Low Median Income</td>
<td>-0.133</td>
<td>0.107</td>
<td>-1.240</td>
<td>0.215</td>
<td>-4.8%</td>
</tr>
<tr>
<td>Transport-Cost Inflation</td>
<td>0.002</td>
<td>0.002</td>
<td>0.950</td>
<td>0.344</td>
<td>3.1%</td>
</tr>
<tr>
<td>Regional Poverty</td>
<td>-0.008</td>
<td>0.011</td>
<td>-0.690</td>
<td>0.488</td>
<td>-2.1%</td>
</tr>
<tr>
<td>Trade Openness</td>
<td>-2.054</td>
<td>0.761</td>
<td>-2.700</td>
<td>0.007</td>
<td>-6.4%</td>
</tr>
<tr>
<td>Strikes</td>
<td>0.206</td>
<td>0.265</td>
<td>0.780</td>
<td>0.437</td>
<td>3.6%</td>
</tr>
<tr>
<td>Inflation Volatility</td>
<td>-0.005</td>
<td>0.003</td>
<td>-1.440</td>
<td>0.149</td>
<td>-4.1%</td>
</tr>
<tr>
<td>National Poverty</td>
<td>-0.214</td>
<td>0.082</td>
<td>-2.620</td>
<td>0.009</td>
<td>-9.6%</td>
</tr>
</tbody>
</table>

-2*ln(L): 290.067

<table>
<thead>
<tr>
<th></th>
<th>Coefficient</th>
<th>Standard Error</th>
<th>t statistic</th>
<th>p value</th>
<th>Quantitative Significance</th>
</tr>
</thead>
<tbody>
<tr>
<td>Constant</td>
<td>15.058</td>
<td>2.524</td>
<td>5.970</td>
<td>0.000</td>
<td>0.0%</td>
</tr>
<tr>
<td>Education</td>
<td>-0.066</td>
<td>0.024</td>
<td>-2.750</td>
<td>0.006</td>
<td>-5.6%</td>
</tr>
<tr>
<td>Initial Income</td>
<td>-0.092</td>
<td>0.038</td>
<td>-2.410</td>
<td>0.016</td>
<td>-5.6%</td>
</tr>
<tr>
<td>Language</td>
<td>--</td>
<td>--</td>
<td>--</td>
<td>--</td>
<td>--</td>
</tr>
<tr>
<td>High Median Income</td>
<td>0.150</td>
<td>0.126</td>
<td>1.190</td>
<td>0.234</td>
<td>4.2%</td>
</tr>
<tr>
<td>Low Median Income</td>
<td>-0.149</td>
<td>0.086</td>
<td>-1.730</td>
<td>0.084</td>
<td>-5.5%</td>
</tr>
<tr>
<td>Transport-Cost Inflation</td>
<td>--</td>
<td>--</td>
<td>--</td>
<td>--</td>
<td>--</td>
</tr>
<tr>
<td>Regional Poverty</td>
<td>--</td>
<td>--</td>
<td>--</td>
<td>--</td>
<td>--</td>
</tr>
<tr>
<td>Trade Openness</td>
<td>-2.299</td>
<td>0.695</td>
<td>-3.310</td>
<td>0.001</td>
<td>-7.2%</td>
</tr>
<tr>
<td>Strikes</td>
<td>--</td>
<td>--</td>
<td>--</td>
<td>--</td>
<td>--</td>
</tr>
<tr>
<td>Inflation Volatility</td>
<td>-0.004</td>
<td>0.003</td>
<td>-1.250</td>
<td>0.212</td>
<td>-3.5%</td>
</tr>
<tr>
<td>National Poverty</td>
<td>-0.253</td>
<td>0.053</td>
<td>-4.820</td>
<td>0.000</td>
<td>-11.4%</td>
</tr>
</tbody>
</table>

-2*ln(L): 292.400

Notes: Quantitative significance indicates the impact on the probability of a region being defined as economically integrated of a one-standard-deviation increase in the indicated variable. Regarding fit, count R² statistics are 0.83 in both cases. The exclusion of Language, Transport-Cost Inflation, Regional Poverty and Strikes from the model yields a likelihood ratio statistic of 2.33, which has a p value of 0.67.
## Table 2 – Misclassification Model

<table>
<thead>
<tr>
<th></th>
<th>Coefficient</th>
<th>Standard Error</th>
<th>t statistic</th>
<th>p value</th>
<th>Quantitative Significance</th>
</tr>
</thead>
<tbody>
<tr>
<td>Pr(y=1</td>
<td>y*=0)</td>
<td>0.003</td>
<td>0.561</td>
<td>0.006</td>
<td>0.498</td>
</tr>
<tr>
<td>Pr(y=0</td>
<td>y*=1)</td>
<td>0.069</td>
<td>0.039</td>
<td>1.752</td>
<td>0.040</td>
</tr>
<tr>
<td>Constant</td>
<td>23.873</td>
<td>12.806</td>
<td>1.864</td>
<td>0.031</td>
<td>0.0%</td>
</tr>
<tr>
<td>Education</td>
<td>-0.121</td>
<td>0.079</td>
<td>-1.53</td>
<td>0.063</td>
<td>-4.3%</td>
</tr>
<tr>
<td>Initial Income</td>
<td>-0.168</td>
<td>0.112</td>
<td>-1.505</td>
<td>0.066</td>
<td>-4.2%</td>
</tr>
<tr>
<td>High Median Income</td>
<td>0.503</td>
<td>0.361</td>
<td>1.394</td>
<td>0.082</td>
<td>5.8%</td>
</tr>
<tr>
<td>Low Median Income</td>
<td>-0.207</td>
<td>0.166</td>
<td>-1.248</td>
<td>0.106</td>
<td>-3.2%</td>
</tr>
<tr>
<td>Trade Openness</td>
<td>-4.197</td>
<td>2.732</td>
<td>-1.536</td>
<td>0.062</td>
<td>-5.5%</td>
</tr>
<tr>
<td>Inflation Volatility</td>
<td>-0.006</td>
<td>0.007</td>
<td>-0.761</td>
<td>0.223</td>
<td>-2.0%</td>
</tr>
<tr>
<td>National Poverty</td>
<td>-0.333</td>
<td>0.183</td>
<td>-1.818</td>
<td>0.035</td>
<td>-6.2%</td>
</tr>
</tbody>
</table>

Notes: Pr(y=1|y*=0) indicates the probability of a non-integrated region being misclassified as an integrated region, and Pr(y=0|y*=1) indicates the probability of an integrated region being misclassified as a non-integrated region.
### Table 3 – OLS Estimates

<table>
<thead>
<tr>
<th></th>
<th>Coefficient</th>
<th>Standard Error</th>
<th>t statistic</th>
<th>p value</th>
<th>Quantitative Significance</th>
</tr>
</thead>
<tbody>
<tr>
<td>Constant</td>
<td>13.957</td>
<td>4.629</td>
<td>3.015</td>
<td>0.001</td>
<td>0.000</td>
</tr>
<tr>
<td>Education</td>
<td>-0.041</td>
<td>0.024</td>
<td>-1.714</td>
<td>0.044</td>
<td>-0.150</td>
</tr>
<tr>
<td>Initial Income</td>
<td>-0.156</td>
<td>0.042</td>
<td>-3.722</td>
<td>0.000</td>
<td>-0.401</td>
</tr>
<tr>
<td>Language</td>
<td>0.010</td>
<td>0.007</td>
<td>1.453</td>
<td>0.073</td>
<td>0.147</td>
</tr>
<tr>
<td>High Median Income</td>
<td>0.279</td>
<td>0.113</td>
<td>2.469</td>
<td>0.007</td>
<td>0.330</td>
</tr>
<tr>
<td>Low Median Income</td>
<td>0.154</td>
<td>0.098</td>
<td>1.570</td>
<td>0.059</td>
<td>0.240</td>
</tr>
<tr>
<td>Transport-Cost</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Inflation</td>
<td>0.001</td>
<td>0.002</td>
<td>0.462</td>
<td>0.322</td>
<td>0.061</td>
</tr>
<tr>
<td>Regional Poverty</td>
<td>-0.016</td>
<td>0.013</td>
<td>-1.298</td>
<td>0.098</td>
<td>-0.192</td>
</tr>
<tr>
<td>Trade Openness</td>
<td>-3.162</td>
<td>0.813</td>
<td>-3.892</td>
<td>0.000</td>
<td>-0.424</td>
</tr>
<tr>
<td>Strikes</td>
<td>0.608</td>
<td>0.266</td>
<td>2.282</td>
<td>0.012</td>
<td>0.455</td>
</tr>
<tr>
<td>Inflation Volatility</td>
<td>-0.000</td>
<td>0.003</td>
<td>-0.070</td>
<td>0.472</td>
<td>-0.008</td>
</tr>
<tr>
<td>National Poverty</td>
<td>-0.189</td>
<td>0.086</td>
<td>-2.202</td>
<td>0.014</td>
<td>-0.362</td>
</tr>
</tbody>
</table>

$R^2 : 0.255$

Notes: Quantitative significance indicates the impact of a one-standard-deviation increase in the indicated variable. Standard errors are heteroskedasticity-consistent (White, 1980).
Figure 1. Aggregate Integration Measure

Percentage of Integrated Regions