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ABSTRACT

Rural Windfall or a New Resource Curse? Coca, Income, and Civil Conflict in Colombia*

Natural and agricultural resources for which there is a substantial black market, such as coca, opium, and diamonds, appear especially likely to be exploited by the parties to a civil conflict. Even legally traded commodities such as oil and timber have been linked to civil war. On the other hand, these resources may also provide one of the few reliable sources of income in the countryside. In this paper, we study the economic and social consequences of a major exogenous shift in the production of one such resource – coca paste – into Colombia, where most coca leaf is now harvested. Our analysis shows that this shift generated only modest economic gains in rural areas, primarily in the form of increased self-employment earnings and increased labor supply by teenage boys. The results also suggest that the rural areas which saw accelerated coca production subsequently became more violent, while urban areas were affected little. The acceleration in violence is greater in departments (provinces) where there was a pre-coca guerilla presence. Taken together, these findings are consistent with the view that the Colombian civil conflict is fueled by the financial opportunities that coca provides, and that the consequent rent-seeking activity by combatants limits the economic gains from coca cultivation.

JEL Classification: Q34, O13

Keywords: rural development, economic shocks, civil war, illegal drugs

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"If it weren't for the armed groups, I think we could reach a consensus on what the region needs to progress. But all the armed groups want is to control the economic question, and all are willing to massacre or murder or force people from their homes to win."

-- Gloria Cuartas, major of Apartadó (quoted in Kirk, 2003).

I. Introduction

Nowhere is the interest in regional economic conditions more acute than in war-torn nations or regions embroiled in civil conflict. Perhaps not coincidentally, many such areas appear to have local economies that depend to a large extent on natural resources, especially those related to illegal economic activities or products for which there is a black market. Examples include the drug trade in Latin America and Afghanistan and the so-called blood diamonds in Africa. Even legal extraction activities, such as timber harvesting and oil mining, have been associated with social breakdown (Ross, 2001). The concentration of extraction activities in conflict zones raises the question of whether this association is causal. Although an increase in resource income may reduce poverty, thereby moderating combatants' desire to fight, natural resources also give the parties to a conflict something to fight over. Moreover, the income from resources provides financing for continued conflict.

The idea that resource wealth can be bad for development is sometimes known as the "resource curse" (e.g., Sachs and Warner, 2000). Economic analyses of the resource curse typically focus on the possibilities of an export-induced Dutch disease and effects on government corruption or rent seeking (e.g., Sala-i-Martin and Subramanian, 2003; Hausmann and Rigobon, 2003). The effect of natural resources on the incidence and duration of civil wars provides a less-explored channel by which natural resources may have perverse effects. This channel features in a burgeoning political science literature, which includes empirical contributions by Collier, Hoeffler and Soderbom (2004), Fearon (2004), and Ross (2003). An antecedent in economics is the theoretical analysis by Grossman (1991). There is also some circumstantial evidence that illegal resources such as drugs increase the duration of civil conflicts (Ross, 2004), but economists and political scientists have yet to produce evidence on this question from a compelling natural experiment.

In this paper we use a quasi-experimental research design to study the impact of demand shocks for illicit resources on rural economic conditions and civil conflict. The setting for our study is Colombia, a good laboratory for our study since almost all of the cocaine consumed in North America and Europe comes from the Andean nations of Bolivia, Colombia, and Peru (United Nations, 2001). Moreover, we exploit a sharp change in the structure of the Andean drug industry: before 1994, most of the cocaine exported from Colombia was refined from coca leaf grown in Bolivia and Peru. Beginning in 1994, however, in response to increasingly effective air interdiction by American and local militaries, the so-called air bridge that ferried coca paste from growers to Colombian refiners was disrupted. In response, coca cultivation and paste production shifted to Colombia's countryside, where it eventually surpassed pre-interdiction levels. We use this shift in an attempt to assess the consequences of the coca economy for Colombia's rural population.

The first question considered here is whether increased demand for coca affected economic conditions for the rural population in ways we can measure using survey data. In particular, the end of the air bridge provides a mirror which can be used to look at the claim that drug interdiction has substantial economic costs for rural producers (see, e.g., Leons, 1997, and Chauvin, 1999). If interdiction is costly, then the post-air-bridge Colombian coca boom of the early 1990s should have had substantial economic benefits. We therefore look at effects on earnings, labor supply, and income, as well as child labor and school enrollment. Of course, coca cultivation *per se* may do little to enrich the cultivators, since – as with the relationship between the farmgate price of coffee and the beans we buy at Starbucks – the price of raw coca leaf makes up a small fraction of the price of cocaine (Alvarez, 1995). On the other hand, most estimates suggest cocaine plays a large enough role in the Colombian economy for changes in the demand for coca to have a perceptible economic effect.

The widely observed association between illicit crops and civil strife raises the related question of whether an increase in coca cultivation has an impact on violence by increasing the resources available to insurgent groups. The link with violence is especially relevant in Colombia, which has experienced some

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¹ For example, Steiner (1998) estimates total Colombian income from illegal drugs at 4-6 percent of GDP in the first half of the 1990s. See also Thoumi (2002).

of the highest homicide rates in the world. This is in spite of substantial economic growth through most of the 20th Century and Colombia's status as one of the oldest democracies in Latin America. The effect of the drug trade on violence has been widely debated in Colombian policy circles (see, e.g., Cardenas, 2001). While a link at first seems obvious, it bears emphasizing that the historical record is ambiguous. Marijuana became an important crop only in the 1960s and the cocaine trade began in the 1970s, with significant coca plantings appearing only in the 1990s (see, e.g., Bagley, 1988). Yet violence and civil conflict, especially outside the major cities, have been a major factor in Colombian political life since independence. During the period known as 'La Violencia' (1948-57), as many as 200,000 Colombians were killed (Winn, 1999). Clearly, cocaine cannot be blamed for starting this conflict, though it may play a role in perpetuating it.

Weighing in favor of a link between the Colombian drug trade and violence is the fact that some of the more recent violence is the work of drug cartels or individuals operating on their behalf. Thus, homicide rates peaked in the late 1980s and early 1990s, when the cartel leadership rebelled against extradition efforts. Probably more importantly, the major Colombian Guerilla groups, especially the Colombian Revolutionary Armed Forces (FARC) and the National Liberation Army (ELN), are widely believed to derive substantial income by taxing drug proceeds, as do illegal self-defense groups or paramilitaries (Rangel, 2000; Rabasa and Chalk, 2001; Villalon, 2004).²

Although the evidence is not seamless, two broad features of our findings tend to support the view that coca fuels Colombia's seemingly interminable civil conflict, while generating few economic benefits for local residents. First, in contrast with the Black, McKinnish, and Sanders (2005) study of coal-mining regions in the US, coca does not boost earnings in an entire growing region, though it is

² Guerilla and paramilitary groups do not refer to control of the drug trade as a primary goal of the conflict. Rather, the rhetoric on both sides concerns security for various constituencies and social justice. For instance, the FARC was initially a farmers' defense coalition which formed in the 1950s to resist the minority conservative government. Later, the FARC established ties with the Colombian Communist party. The ELN was created by university students and inspired by Che Guevara and the Cuban revolution. The AUC ostensibly protects the interests of ranchers, farmers, and other landowners. The income from taxing drug proceeds appears to fund political action, maintenance of a precarious social security system for members and their families, occasional work on infrastructure, and combat activity including weapons purchases (Rangel, 2000).

associated with increased self-employment income for those already active in this sector. This is consistent with anecdotal evidence that the economic benefits of coca growing are largely taxed away by combatants or otherwise dissipated through non-productive activities. Second, in spite of the fact that income and hours worked increased for some groups, violence also increased in regions where coca cultivation increased. This runs counter to the findings in Miguel, Satyanath, and Sergenti (2004), who link improvements in economic conditions generated by rainfall to decreased civil conflict in Africa, but appears consistent with economic theories of rent-seeking behavior by combatants (e.g., Grossman, 1991; Collier and Hoeffler, 2004).³

The paper is organized as follows. The next section provides additional background. Section III outlines the approach we used to divide Colombia into coca-growing and non-growing regions for the purposes of our within-country survey-based analysis. Section IV discusses estimates of the effect of coca growing on rural economic conditions and Section V presents the mortality estimates. Section VI summarizes and interprets the results.

II. Institutional Background and Economic Framework

Until the early 1990s, coca was mainly harvested in Bolivia and Peru, after which most cultivation moved to Colombia.⁴ Whether in Bolivia, Peru, or Colombia, Coca is typically grown in thousands of small peasant holdings. Harvested coca leaves are dried by farmers and sold to entrepreneurs who make them into coca paste, a simple chemical process that takes a few days. Paste has about one-hundredth the volume of coca leaves, and the transition from leaf to paste is where most of the weight reduction in cocaine production occurs. The next step in coca processing is to make coca base, a somewhat more complicated chemical process. Finally, cocaine hydrochloride is refined from coca base,

³ Diaz and Sanchez (2004) offer a recent exploration of the coca-conflict nexus in Colombia, arguing that conflict causes coca and not vice versa, but their spatial-correlations research design does not exploit exogenous shifts. In related studies, Guidolin and La Ferrara (2005) and Pshisva and Suarez (2004) put conflict and kidnappings on the right-hand-side of a firm-level investment equation. A number of studies have also used cross-country data to address the association between social conflict, institutions, and growth (e.g., Rodrik (1999)), and natural resources, institutions, and growth (Mehlum, Moene, and Torvik (2006)).

⁴ This section draws on Whynes (1992) and Thoumi (1995).

a chemical process that often occurs in towns or cities. Street cocaine is made by diluting cocaine hydrochloride with sugar and baking soda, usually in the consuming country.

While Colombia has almost always been the principal exporter of refined cocaine, until fairly recently little coca was grown there. Colombian middlemen and exporters operated by importing coca paste (or coca base) from Bolivia and Peru, specializing in refining and distributing cocaine hydrochloride (i.e., cocaine). In the early 1990s, the drug industry changed in response to a change in emphasis in US and producer-country enforcement policies. In April 1992, after Peruvian president Fujimori's so-called self-coup, the Peruvian military began aggressively targeting jungle airstrips and small planes suspected of carrying coca paste, as part of a general process of militarization of the drug war (Zirnite, 1998). Colombia followed suit in 1994 with a similar shoot-down policy for planes ferrying paste from both Peru and Bolivia. US policy moved in tandem with Presidential Decision Directive 14 in November 1993, which shifted U.S. interdiction away from Caribbean transit zones like Bermuda towards an attempt to stop cocaine production in the Andes. The disruption of the air bridge ferrying coca paste into Colombia was a key part of this effort.⁵

The militarization of the drug war and disruption of the air bridge does not appear to have reduced the supply of cocaine (see, e.g., Rabasa and Chalk, 2001). It did, however, lead to a marked shift in the organization of the industry among producer countries. This can be seen in Figure 1, which uses data from a United Nations (2001) drug report to show the change in the locus of production of dry coca leaf from Peru and Bolivia to Colombia. While Bolivian production was flat in the early 1990s, it started to decline in 1994. In contrast, Peruvian production fell sharply from 1992 to 1993 and again in 1995, followed by a sharp and steady increase in Colombian production from 1993-94 and continuing thereafter. Part of this increase appears to have come from increased cultivation and part from improved yields. Colombian production continued to grow thereafter, as did the Colombian share of total

⁵ The Peruvian and Colombian shoot-down policies can be seen as a response to U.S. pressure. Militarization of the drug war began as part of first President Bush's "Andean Strategy" in 1990, with a program of military, economic, and law-enforcement assistance for Andean nations in FY1990-94. Initially, however, this effort met with little sympathy in the region (Washington Office on Latin America, 1991). Late 1992 and 1993 marked the beginning of a period of independent efforts and sharply increased cooperation by producer nations.

production. Other figures in United Nations (2001) show that by 1997, potential coca production in Colombia (i.e., before crop eradication following Plan Colombia in 2001) exceeded that in Peru.

Economic framework

We see the end of the air bridge as initiating an exogenous fall in the price of coca (leaf, paste, or base) in the traditional producer nations of Bolivia and Peru, while causing a price increase in Colombia. The price in traditional growing countries fell when coca could no longer be shipped to Colombian refineries and distributors. Peruvian and Bolivian growers have no competitive export channels of their own since they have no Caribbean ports and because their foreign distribution networks are not well-developed. At the same time, the price of coca grown in Colombia increased when drug middlemen and entrepreneurs tried to elicit new and more accessible supplies. Farmers and potential farmers responded to the increase in the price of coca by growing more of it, a response that very likely accounts for the pattern in Figure 1, though non-economic factors may have been at work as well.

Did the end of the air bridge really change coca prices in the manner described above? Although we do not have a reliable time series of coca prices by producer country, anecdotal evidence supports this description of the coca market in the mid-1990s. For example, Zirnite (1998, p. 171) quotes the regional US military commander testifying to Congress in 1996 that, "the so-called air bridge between Peru and Colombia saw a greater than 50% temporary reduction of flights," and that consequently, ". . . there was a glut of coca base on the market and the price of the product being shipped fell 50 percent overall and by as much as 80 percent in some areas." On the Colombian side, data reported in Uribe (1997, p. 62) for the department of Guaviare show the price of base more than doubled from 1992-94. Journalistic accounts similarly point to an increase in prices in Colombia (e.g., Villalon, 2004).

This description suggests a number of channels through which increased coca cultivation might

⁶ A related question is why coca was not previously grown in large quantities in Colombia. The answer appears to be that Colombian coca farms were less productive; see p. 71 in Uribe (1997). Consistent with the increase in coca production, the production of Colombian coffee, which like coca is grown mainly in small plots, turned sharply downwards in the mid-late 1990s, after increasing over most of the previous two decades (see http://www.dane.gov.co/inf est/ena.html).

affect economic conditions and the level of rural violence in coca-growing regions. The increase in coca prices presumably made coca farmers better off, with possible aggregate regional effects of the sort documented by Black, McKinnish, and Sanders (2005) in the Appalachian coal-mining region and by Carrington (1996) in Alaska. Increased prices for coca production generated new sources of revenue for taxation. Because the central government is weak in the Colombian countryside, these opportunities most likely benefited guerillas and paramilitaries. Of course, if coca taxes are too high, then there is no incentive to produce. Taxes were imposed not only at the point of sale, however, but also through kidnapping, extortion, and based on the guerilla's "economic census", a sort of partisan's tax return (Rangel, 2000, p. 588).⁷ This tax and extortion system may have transferred a large fraction of the economic benefits of coca production to combatants, while still leaving coca production more attractive than alternative activities. In addition, to the extent that coca finances a disruptive civil conflict, increased coca production may have reduced the overall level of economic activity.⁸

III. Classification of Regions

Our research design exploits the fact that the change in the drug industry in the early 1990s probably had a disproportionate effect on Colombian departments which, by virtue of climate and soil conditions, politics, or infrastructure were hospitable to the cultivation of coca plants and the production of coca paste. This naturally raises the question of how to classify departments or regions as potential coca-growers and paste-producers. The best candidates for future coca production seem likely to be departments with a pre-existing coca presence. We identified baseline coca-growing departments using estimates for 1994 reported in Uribe (1997, p. 67). This source collects a number of international

⁷ About 85% and 65% of the FARC's and ELN's revenues, respectively, are estimated to come from drugs and extortion (Rangel, 2000, p. 585). While similar estimates of revenue sources for paramilitary groups do not exist, these groups are also widely believed to benefit from the drug trade. Grossman and Mejia (2004) develop a theoretical model of guerilla involvement in drug production.

A caveat here, relevant for our empirical strategy, is the possibility of general equilibrium effects in non-growing regions. Examples include price effects and migration of labor. Although we cannot look at regional price variation, we found no evidence of coca-related migration patterns at the departmental level. Moreover, in the case of coca, extensive linkages with other consumer prices and farm input prices seem unlikely, given that cocaine is primarily for export and coca production requires few inputs other than labor.

observers' estimates of hectares of coca bush under cultivation in Colombian departments. The reports summarized in the table are dated October 1994, so the data were presumably collected somewhat earlier. The 9 departments that had at least 1,000 hectares under cultivation are Bolivar, Caqueta, Cauca, Meta, Narino, Putumayo, Guaviare, Vaupes, and Vichada.⁹

In a second coding scheme, we expanded the definition of the growing region to include the five additional departments identified as growing on a satellite map in Perafan (1999, p. 11). This map is also dated 1994. The Perafan map adds the three Northern departments of Cesar, Magdalena, and La Guajira, and the departments of Norte de Santander and Guainia. These 5 are also listed as growing regions in Uribe (1997), while all in the group of 9 appear as growing on Perafan's (1999) map. We refer to the expanded coding scheme as defining a "14-department growing region" and the 5 additional departments added to the 9 growing departments to construct this region as "medium producers".

Our color-coded map, reproduced in the Appendix, shows the 9-department growing region to be concentrated in the Southern and Eastern part of the country. Note, however, that not all Southern or Eastern departments grow large amounts of coca. For example, Amazonas, in the Southeast corner, and Arauca, in the East, are not coded as a growing department in either scheme. The group of 9 growing departments includes two, Meta and Caqueta, which were ceded to FARC control from 1998 to 2001 as part of an abortive peace effort. We refer to these two as the demilitarized zone (DMZ) and allow for separate DMZ effects in some of the empirical work. The five departments coded as medium producers are mostly in the Northern part of the country, though one, Guainia, is in the far Eastern region. As a final check on the results, we also distinguish all departments on the basis of previous guerilla activity.

To establish a "first-stage" relation for our division of Colombian departments into growing and non-growing regions, we report the results of a regression of the growth in coca cultivation from 1994 to 1999 or 1994 to 2000 on an indicator for growing status in 1994. Growth is measured from a 1994 base since this is the year used to classify growing regions (as noted earlier, the 1994 data were probably collected earlier). The endpoint years of 1999 and 2000 are used because these are the years for which

⁹ Black, McKinnish, and Sanders (2002) similarly identify counties affected by the coal boom and bust using preexisting production data.

departmental cultivation figures are available. In any case, the change from 1994 to the end of the decade seems likely to provide a good summary of coca penetration in the relevant period.¹⁰

The first-stage results, summarized in Table 1, show a strong correlation between coca growth and base-period growing status.¹¹ The estimates in Column (1), Panel A indicate that cultivation grew by about 7,500 more hectares in the 9-department growing region than elsewhere, while the omission of medium producers leads to a slightly larger effect. Omission of the two DMZ departments leads to an even larger effect of almost 9,000 hectares, shown in Column (2) of Panel B. With or without DMZ departments, the growth effect is significantly different from zero. The estimates in Columns (5)-(8) also show mostly larger effects when growth is measured through 2000 instead of 1999, with growing regions gaining 8,961 (s.e.=4,358) hectares over the period in the sample without medium producers.¹² None of the intercept estimates are significantly different from zero, indicating essentially no growth in the departments with no initial production in 1994. Finally, estimates with growing status defined using the 14-department scheme, i.e., moving the medium producers to the treated group, also show substantial growth in cultivation, but less than in the 9-department subset omitting medium producers. The 14-department scheme also generates a smaller intercept.

An interesting finding in this context, relevant for our choice of estimation strategy, is that dummies for the two growing regions do a better job of predicting coca growth than a linear predictor using base-period levels. Results from the linear parameterization can be seen in the last two rows of each panel of Table 1. A visual representation of alternate parameterizations is presented in Figure 2,

¹⁰ The 1999-2000 data are from Colombia's anti-drug agency, Direccion National de Estupefacientes (DNE, 2002), collected through the Illicit Crop Monitoring System (SIMCI- Sistema Integrado de Monitoreo de Cultivos Ilicitos). This system was implemented by the United Nations Office on Drugs and Crime with the logistical support of the Colombian anti-narcotics Police (DIRAN) and in coordination with the DNE. The data are from satellite images and verification flights. Data for 2000 appear to be more complete than the 1999 data.

¹¹ Our use of the term "first-stage" in this context is motivated by the fact that, given consistent departmental time series data on coca production, we could use interactions between initial growing conditions and a post-air-bridge dummy as an instrumental variable for the effects of endogenous coca production on economic conditions and violence. In the absence of reliable data on the relevant endogenous variable, we focus below on the reduced form regressions of economic and mortality outcomes on initial conditions/time interactions.

¹² Mean growth was about 2,800 hectares through 1999 and 2,900 through 2000. The 1994 mean for hectares under cultivation is about 2,100. The base mean was 7,155 in the 9-department growing region and 4,732 in the 14-department growing region. We estimate that in 1994, roughly 15-19 percent of cultivated hectares were devoted to coca in the 14- and 9-department growing regions.

which plots coca growth against base period levels, using different symbols for the non-growing region, the 9-department growing region, and the remaining growing region on a log scale. The two growing regions have much higher coca growth, but the relationship between base period levels and growth rates is not especially linear. Although the best single predictor of coca growth is a dummy for the 9-department region, the empirical work below also uses the 14-region scheme since this turns out to balance pre-treatment homicide rates better than the 9-department scheme and because the rural household survey is missing some growing departments.¹³

It is also worth emphasizing that the empirical first stage is not meant to provide precise measure of the link between base-period levels and the growth in coca cultivation. If information on coca cultivation is subject to transitory classical measurement error, then the first-stage estimates reported here underestimate the impact of levels on growth. Moreover, producing regions may do a better job of hiding cultivated areas, leading to errors in satellite data that are negatively correlated with levels, further exacerbating attenuation bias. This suggests the first-stage estimates should be viewed as an underestimate, and may explain why categorical variables do a better job than linear terms at predicting cultivation growth. The division into growing categories is based on an average of rough estimates for 1994 from three different observers and sources. These data seem likely to capture the distinction between areas with substantial coca cultivation and areas with little or none even if the 1999-2000 satellite data are noisy.

Descriptive Statistics by Region Type

Not surprisingly, the growing departments are more rural and somewhat poorer than the rest of the country. This is apparent in the descriptive statistics in Table 2, which compares growing and non-

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¹³ The ratio of coca hectares to non-coca hectares under cultivation grew by .33 from 1994-99 and by .99 from 1994-2000 in the 9-department growing region. The corresponding statistics for the 14-department region are .18 and .62. Like the figures in levels, these are rough estimates, but they serve to identify regions with a strong and growing coca presence (proportional coca cultivation declined slightly in the non-growing region, by about .06). We also attempted to define growing regions based on climate and soil conditions using geographic information from Sánchez and Núñez (2000). In practice, this does not produce as strong a first stage as a classification schemes based on 1994 levels, probably because coca grows under a broad range of conditions (Thoumi, 2002, p.105).

growing regions along a number of dimensions. The comparison between growing and non-growing regions is affected by the fact that the non-growing region includes the three departments with Colombia's largest cities: the Bogota capital district; Antioquia, which contains Medellin, an especially violent city; and Valle del Cauca, where Cali is located. To improve comparability with growing regions when comparing homicide rates, we tabulated statistics without these 3 departments. The Bogota capital district, Antioquia and Valle del Cauca are also dropped from the mortality analyses in order to avoid confounding with the secular decline in violence in big cities in the early nineties. Only the Bogota capital district is dropped from the analysis of rural labor markets and rural income.

Omitting the 3 big-city departments, the non-growing population is 65 percent urban, in comparison to 50 percent urban in the 9 department region minus the DMZ, 58 percent in the DMZ, and 66 percent urban in the 5 additional growing departments (medium producers). Although growing and non-growing departments differ along the urban/rural dimension, they had similar primary school enrollment rates. Secondary school enrollment was somewhat lower in the growing regions, consistent with the fact that these regions are more rural, and have lower per-capita GSP. On the other hand, without the three big-city departments, the contrast in income levels between the growing and non-growing regions is considerably reduced.

Homicide rates in the early 1990s were unusually high, even by Colombian standards. For example, the homicide rate reached a remarkable 719 per 100,000 in Antioquia, mostly because of violence in Medellin, and was 272 overall in the non-growing region. These statistics are per 100,000, among men aged 15-59. Without the big-city departments, homicide rates in the non-growing region averaged 141 per 100,000. This can be compared to the rates of 87 in the 9-department growing region without the DMZ, 151 in the medium producers, and 205 in the DMZ. Thus, omission of big-city departments makes homicide rates somewhat more comparable across regions. Our working paper (Angrist and Kugler, 2005) discusses these homicide rates in a broader Latin American context.

Potential Confounding Factors

A potential complication for our analysis is the fact that many growing departments were previously centers of guerilla activity. We would therefore like to distinguish growth in insurgent activity due to coca from a secular expansion in areas where central government control was already weak. In an attempt to distinguish coca-induced effects from the direct effect of a strong guerilla presence, we estimate some models allowing for separate *trends* in regions (whether growing or not) with substantial pre-existing guerilla activity. On the other hand, *treatment effects* on violence might also be expected to be larger in areas where guerillas already had a foothold, since a well-established guerilla movement may be especially likely to benefit from resources from illegal activities. This possibility is therefore explored in a subset of the analyses as well.

A second consideration in the Colombian context is the large number of economic migrants who move to rural areas in search of work and especially the flow of refugees out of the countryside as a consequence of the civil conflict ("poblacion desplazada"). Both types of migration may induce selection bias in an analysis of economic circumstances in rural areas. As a partial check on this, we report results from samples with and without migrants. It is also worth noting that much war-related displacement occurs within departments, and that, according to United Nations High Commission for Refugees (UNHCR, 2002), the largest senders and receivers of displaced populations include both growing and non-growing departments under our classification scheme. In addition, the phenomenon of internal displacement long pre-dates the rise in coca production. In fact, a specification check which looks for growing-region/year interactions of the sort that might confound our analysis shows no growing/post-1995 effect on the probability of being a migrant.

Finally, a series of economic reforms since 1990 may be relevant. In 1991, the government of President Gaviria introduced a sharp reduction in tariffs and undertook steps towards deregulation of labor and financial markets (these are discussed in Kugler, 1999 and Eslava *et al*, 2004). In 1993, Gaviria's government also introduced a social security reform (Kugler and Kugler, 2003; Kugler, 2005). Around 1997, President Samper's government introduced a minor tax reform and privatized the energy

sector. However, these structural reforms were adopted at the national level and should have affected growing and non-growing regions alike. In principal, a decentralization effort in 1991 may have affected different types of regions differently. In practice, however, this change was very limited, with tax collection and most spending remaining under central control (Alesina, 2005). Finally, Plan Colombia, an important American-sponsored aid and anti-drug initiative came on the scene after our period of study.

IV. The Economic Consequences of a Coca Economy

Data and Descriptive Statistics

This section uses differences-in-differences type regressions to assess the economic consequences of the shift in coca production to Colombian growing regions. The data come from the rural component of Colombia's annual household survey and are described in the Appendix. The rural survey provides large repeated cross-sections, with information on households and individual household members, including children. We limit the analysis to data from 1992 (because of earlier changes in survey design) through 2000 (after which the survey was replaced by a new panel data set). The survey was conducted in 23 of Colombia's 33 departments. Using the 14-department definition of the growing region, the rural survey includes households from 7 growing departments plus the two DMZ departments. Because only 3 non-DMZ departments from the 9-department growing region were included in the rural survey, we focus initially on the 14-department classification scheme.¹⁴

Our analysis looks separately at samples of adults, school-age children, and teenage boys who might be in the labor market. The sample of adults includes men and women aged 21-59, and is described in the first two columns of Table 3 using data for 1992 and 1997. Roughly 30% of respondents in this sample were migrants, where migrants are defined as individuals who do not currently live in the county where they were born. Most were married and about half are male. The growing region contributed from 24 percent of the sample in 1992 to 30 percent of the sample in 1997. The number of

¹⁴ The included growing departments are Bolivar, Cauca, Narino, Cesar, La Guajira, Magdalena, and Norte de Santander, plus Caqueta and Meta in the DMZ. In contrast with the mortality analysis, discussed below, Antioquia and Valle de Cauca are included in the household analysis because the survey is limited to rural households.

respondents from the DMZ also increased, from 1.4 to 3.9 percent.

About two-thirds of adults in the survey were employed in 1992 and 1997, though only about 36 percent had positive wage and salary earnings. Employment rates for men were 93-95 percent, as can be seen in Columns (3) and (4), and 55 percent of men had positive wage and salary earnings. Between 25 and 26 percent of adult men and women had positive income from self-employment, while between 35 and 37 percent of adult men had positive income from self-employment. Self-employment income includes income from individual short-term contracts, from the sale of domestically produced goods, and from commercial or family-based agricultural production. Wage and salary earnings and self-employment income are reported in real terms and were constructed using the consumer price index provided by the Department of National Statistics (DANE). These variables are given in 1998 pesos, worth about 1,400 to the US dollar. Thus, mean wages range from 52 to 58 dollars per month, and mean self-employment income from 241 to 252 dollars per year, in the sample of adults.

Descriptive statistics for the sample of children, reported in Columns (5)-(8), show that most were enrolled, and enrollment rates increased somewhat between 1992 and 1997. Fewer children than adults were migrants, but the regional distribution of children was broadly similar to that for adults. Employment statistics for children are only collected for those over 10 years of age. About a third of boys aged 10-16 and 10 percent of girls aged 10-16 were working, indicating the importance of child labor. The statistics in Columns (9) and (10) show that over half of boys aged 13-20 were working. Hours per month for boys were substantial, though lower than for adults. Boys also had lower earnings. The wage and salary income of boys ranges from 42-44 dollars per month, and boys' self-employment income ranges from 44-47 dollars per year. Less than half were still in school and few were married.

Results for Adults

The basic empirical framework looks for growing-region/post-air-bridge interactions while controlling for department and year effects. In particular, we estimated year-region interaction terms using the following model for respondent i in department j in year t:

$$y_{ijt} = X_i' \mu + \beta_i + \delta_t + \Gamma_s \alpha_{0s} g_{is} + \Gamma_s \alpha_{1s} d_{is} + \varepsilon_{ijt}, \tag{1}$$

where β_j is a department effect, δ_t is a year effect, g_{js} indicates non-DMZ growing departments when t=s, and d_{js} indicates DMZ departments when t=s (s=1994, . . ., 2000). The parameters α_{0s} and α_{1s} are the corresponding region-type/year interaction terms. Some models also include linear trends for each department type as a control for omitted variables and serial correlation. This amounts to replacing β_i with $\beta_{0j}+\beta_{1j}t$, where β_{1j} takes on 3 values (non-growing, growing, and DMZ). The estimating equations also control for a vector of individual covariates, Xi, which includes sex, age dummies, household size, marital status and migrant status. For binary dependent variables, the linear model was replaced with the analogous logit. Standard errors here and elsewhere are clustered at the department level to allow for correlation across individuals within a state and within states over time. 15

The analysis of rural outcomes begins with estimates of effects on the probability of having selfemployment income and on the log of self-employment income for those who have some. Because coca production is an agricultural activity, self-employment status (either as farmer, employer, landowner, or contractor) is of special interest. The interpretation of results for log self-employment income is potentially complicated by selection bias from conditioning on having earnings in this sector. As in a wage equation, however, we can make an educated guess as to the likely sign of any selection bias. Since the presumptive effect of being in the growing region after 1994 is to increase the likelihood of selfemployment, the conditional-on-positive estimates of effects on log wages will typically be biased downwards by the fact that, on the margin, those induced to enter self-employment have lower selfemployment earnings potential in the absence of treatment (see, e.g., Angrist, 2001).

The first two columns of Table 4a report marginal effects from the logit version of equation (1), with a dummy for self-employment status on the left-hand side. The sample includes women as well as men because women have a reasonably high probability of having self-employment income. The estimates in Column (1) are small, with positive but insignificant effects in 1995-97 and 1999-00. We also report results without migrants as a partial control for potential selection biases from migration into

¹⁵ Standard errors estimated with department-year clustering are similar.

and out of growing regions. Results omitting migrants, reported in Column (2), are somewhat larger, peaking in 1996 at 0.048, with a similar marginally significant effect of 0.051 in 2000 (s.e.=0.029).

In contrast with the small-to-zero estimates for self-employment probabilities, the estimates in Columns (3) and (4) show a substantial increase in (log) self-employment income. In particular, there are large, statistically significant effects on the order of 0.3-0.4 in 1996-98, a period when coca is likely to have had a major impact. For example, the effect in 1996 in the sample including migrants is 0.362 (s.e.=0.155). There are somewhat smaller positive effects in 1995 and 1999-2000.

In an effort to improve precision, we also estimated models with pooled region-year interaction terms. These models can be written

$$\begin{aligned} y_{ijt} &= X_{i}' \mu + \beta_{j} + \delta_{t} \\ &+ \alpha_{0,95-97} g_{j,95-97} + \alpha_{0,98-00} g_{j,98-00} + \alpha_{1,95-97} d_{j,95-97} + \alpha_{1,98-00} d_{j,98-00} + \epsilon_{ijt}, \end{aligned} \tag{2}$$

where X_i is the vector of individual characteristics referred to above, with coefficient vector μ . The interaction dummies $g_{j,95-97}$ and $g_{j,98-00}$ indicate the non-DMZ growing region for t=1995-97 and t=1998-2000, with corresponding interaction terms $\alpha_{0,95-97}$ and $\alpha_{0,98-00}$. Likewise, the interaction dummies $d_{j,95-97}$ and $d_{j,98-00}$ indicate the DMZ in 1995-97 and 1998-2000, with corresponding interaction terms $\alpha_{1,95-97}$ and $\alpha_{1,98-00}$. As before, with binary dependent variables the reported results are logit marginal effects. Also, as with equation (1), we estimated versions of (2) replacing β_j with $\beta_{0j} + \beta_{1j}t$, where β_{0j} is a department fixed effect and β_{1j} is a trend taking on 3 values, one for each department type.

Self-employment results from models with pooled interaction terms and omitting trends are reported in Columns (1) and (3) of Table 4b. These models generate statistically significant estimates of effects on the probability of self-employment and on the log of self-employment income in the non-DMZ growing region. The former effects are small, on the order of 3-4 percentage points, but the latter are large (see, e.g., the Column (3) estimate of 0.25 in 1995-97 with a standard error of 0.121). Moreover, the absence of substantial effects on the probability of having self-employment income suggests selection bias from changes in labor force participation is not much of a concern in this context.

Estimates of interactions terms for the DMZ show no effect on the probability of having self-

employment income, but even larger (though imprecisely estimated) effects on log self-employment income than in the non-DMZ region. Again, these results may be subject to selection bias as a result of migration, especially in the DMZ, though we include a migrant dummy as a partial control. At the same time, as pointed out above, we found no evidence of selection bias due to migration in a regression of the probability of being a migrant on a growing/post-1995 interaction.

The evidence for an effect of the coca boom on the probability of self-employment is weakened considerably by the inclusion of region-specific trends. For example, the estimates reported in Column (2) of Table 4b are either zero or negative. On the other hand, the 1995-97 effect on the log of self-employment income, estimated in a model with region-specific trends, is about the same as when estimated without trends (compare 0.288 and 0.251) and marginally significant (t=1.78). Moreover, the trend itself is not significantly different from zero.

The remaining estimates in Tables 4a and 4b are for effects on labor supply measures and the log of monthly wages in a sample of men. We focus on men because male participation rates are considerably higher than female participation rates, especially in the wage sector. The estimated employment effects for men show little evidence of a change in participation in the growing region. Most of the estimated yearly interaction terms are small and none are significantly different from zero. There is some evidence of an increase in log hours, though it is not very robust. For example, in the hours equation, the 1996 interaction without migrants is 0.048 (s.e.=0.037) and the 1998 interaction with migrants is 0.053 (s.e.=0.023). In models with pooled interactions, there is stronger evidence for a significant effect in 1998-2000 than in 1995-97, though again the estimates are muddied by inclusion of region-specific trends. The pattern of results for log wages similarly offers no robust evidence of an effect.

Results for Children and Youth

We might expect the increase in coca production to have reduced school enrollment and to have

generated an increase in child labor.¹⁶ Columns (1) and (2) in Table 5a indeed show statistically significant reductions of 0.065 and 0.073 in boys' school enrollment in 1997, but estimates for other years are smaller, and none of the corresponding estimates in pooled models, with or without trends, are significant (see Table 5b). Moreover, while the estimated interaction terms without trends are all negative, inclusion of trends causes the signs to flip for boys. Estimates for girls are mainly positive, though on the whole not significant. An exception is the DMZ, where effects are negative and marginally significant without trends.¹⁷

While there appears to have been little impact on school enrollment, the pattern of estimates for teen boys' labor supply is more complex. The one-year employment effect in 1997 and the pooled later-period employment effect in the non-DMZ growing region are positive. On the other hand, the pooled interaction term for the later period is negative and significant for the DMZ, though imprecise and implausibly large in models with trends. The pooled non-DMZ growing effects are also negative when estimated in models with trends. It therefore seems fair to say that there is no robust evidence of an increase in boys' employment rates.

Results for log hours are more clear-cut. In models without trends, log hours appear to have increased in both the non-DMZ growing region and the DMZ. For example, the pooled estimate for the earlier period for the non-DMZ area is 0.112 (s.e.=0.039) and many of the yearly interactions in Table 5a are significant. Inclusion of trends wipes out the DMZ effect but leaves the non-DMZ effects essentially unchanged, though no longer significant. Again, however, the trend in the non-DMZ growing region is insignificant as well. On balance, therefore, Table 5a provides support for the notion that coca production increased teen boys' labor supply, at least in the growing departments outside of the DMZ.

Estimates Using Urban Controls and Without Medium Producers

Although estimates of equations (1) and (2) point to effects on self-employment income for adults

Edmunds and Pavcnik (2004) recently explore the link between trade flows and child labor. Following their taxonomy, coca can be seen as an unskilled-labor intensive good that is a candidate for production with child labor.

¹⁷ To adjust inference for within-household clustering, estimates for children and youth were averaged up to the household level. For details, see the Appendix.

and effects on hours worked by teenage boys, these results are made less precise by the inclusion of region-specific trends. In an effort to increase precision and further check the robustness of these findings, we tried a pooled analysis that stacks urban with rural data for the subset of departments included in both surveys.¹⁸ The idea here is to check whether growing region/post-air-bridge interaction effects are indeed larger in rural than urban parts of growing departments, since we expect income shocks generated by coca to be larger in the countryside. An urban-rural stack also facilitates control for region-specific trends, assuming these trends have similar effects in urban and rural areas. A second modification we explored in an effort to sharpen the growing/non-growing contrast is to drop the 5 medium producer departments from the list of 14 growing regions. This is in the spirit of Black, McKinnish, and Sander's (2005) analysis of coal-producing counties, which also excludes middle-producers according to the level of baseline production.

The estimating equation for the stacked sample allows for urban main effects and urban interactions with both region-type and period dummies in a pooled model similar to the one used to construct the estimates reported in Table 4b. The coefficients of interest are growing-region/post-treatment interaction terms, which are allowed to differ by urban-rural status. This analysis pools growing and DMZ because one DMZ department is missing from the urban survey. Finally, these models control for region-specific trends, which are assumed to be the same in both the rural and urban areas of a given department type. The addition of urban data potentially allows us to estimate these trends more precisely. The stacked analysis is limited to the subset of adult and children outcomes of primary interest and/or for which there appeared to be some evidence of effects in Tables 4 and 5.

Estimates for adult self-employment outcomes in the rural sector, constructed from the urban/rural stack and reported in Columns 1-4 of Table 6, are similar to those generated using rural data only. Again, there is no evidence of an increased likelihood of self-employment in cities or in the countryside (in fact, there is a negative effect for 1998-2000 in urban areas). At the same time, however, the stacked results show a sharp increase in log self-employment earnings for rural workers; in Column

¹⁸ The urban household survey is distinct from the rural survey and has somewhat different geographic coverage and variable definitions. For details, see the Appendix.

(3), for example, the effect is 0.29 (s.e.=0.13) in 1995-97. In contrast, the corresponding urban effect is an insignificant 0.16 (s.e.=0.11). The interaction terms in Column (3) for the 1998-2000 period similarly show larger effects in rural than urban areas. Estimates of effects on log self-employment income using a sample without medium producers are slightly larger than in the full sample, but otherwise similar.

In contrast with the self-employment results, the estimated effects on hours worked by adult men are more mixed. Estimates with medium producers included, reported in Column (5), show no significant rural or urban effects. On the other hand, dropping the medium producers leads to significant rural effects and a smaller but significant urban effect in 1995-97. The rural effect is larger but not significantly different from that in urban areas. Therefore, taking the urban effect as a control for confounding factors, this suggests there is no effect of coca cultivation on adult hours in rural areas. At the same time, Columns (7)-(8) show no effect on wage and salary earnings in either urban or rural locations.

A clearer picture emerges from the analysis of hours worked by teen boys in rural areas. These results, reported in Columns (9)-(10) of Table 6, show mostly significant effects with or without medium producers. The results for teen boys also generate a significant contrast by urban-rural status. In particular, there are substantial increases in hours worked by rural teen boys, with no corresponding effect on teen boys in urban areas. For example, the effect on hours worked in 1995-97 is 0.12 (s.e.=0.052) in rural areas, but -0.004 (s.e.=0.068) in urban areas. Finally, we note that estimates for employment status using the urban/rural stack and corresponding to those in Table 6 (but not reported in this table) show no effects on either adult men or teen boys.

Robustness Checks and Discussion of Magnitudes

The results in Table 6 provide consistent evidence of an increase in self-employment income and hours worked by teen boys in the rural parts of coca-growing departments. As a check on these findings, we estimated a number of variations on the basic set-up using data from the rural sample. These results, reported in Appendix Table A1, are for models with pooled yearly interaction terms and pooled growing and DMZ effects as in Table 6. The first variation adds department rainfall as a time-varying control

variable to the model without department trends; the second includes rainfall and department trends. Neither of these variations has a marked effect on the estimated effects, though as before control for linear trends reduces precision. The remaining two variations allow for different sorts of trends, first an interaction between per-capita gross state (department) product in 1993 and second, a linear trend specific to departments with a substantial FARC presence in 1994. The results here also change little, showing similar increases in self-employment income and teen boys' log hours in the growing regions. ¹⁹

To give a sense of whether the magnitude of self-employment earnings effects in Columns (3)-(4) of Table 6 can plausibly be attributed to a coca boom, we take .2 as a benchmark, a number between the rural effect of 0.29 from Column (3) and the corresponding difference in urban and rural effects, 0.29–0.16=0.13. To calibrate, we use Uribe's (1997) description of a typical family coca farm, consisting of a half-hectare plot that generates about 110,000 pesos/month in revenues from the sale of coca leaf. Our estimates imply an increase in monthly self-employment earnings of about 22,800 pesos at the mean positive self-employment income for self-employed workers (roughly 114,000 pesos per month). We do not know how many self-employed workers were actually growing coca. But assuming a quarter of self-employed adults in rural growing regions had small coca plots, an aggregate increase of 20% in self-employment earnings could have been generated by a 40% increase in output among existing producers jointly with a 40% increase in prices. This is obviously just a rough guess. The point is that the magnitude of the price increase and the earnings from coca at baseline are very likely large enough to sustain the kind of impact suggested by our estimates.

V. Coca and Violence

The estimates in the previous section suggest some localized benefits from the coca boom. The benefits are largely those that might be expected to accrue to farmers or others directly involved in the coca industry, including a marginal labor force of teenage boys. On the other hand, there is little evidence

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¹⁹ The additional data used for robustness checks are documented at the end of the data appendix. The FARC data used to construct the estimates in Table A1 were also used to construct some of the estimates in Table A2 and the estimates in Panel A of Table 10, discussed below.

of wider gains or regional spillovers in the form of a region-wide increase in earnings. In this section, we turn to an analysis of coca effects on violence, as captured by changing homicide rates.

Graphical Analysis

The evolution of violent death rates in the 1990s is described in Figure 3a, which plots death rates per 100,000 for men aged 15-59 by region type, after removing group means. This figure pools the DMZ with other growing departments defined using the 14-department scheme. The resulting plot shows a remarkably parallel evolution of violent death rates in the growing and non-growing areas through 1993. In particular, the growing and non-growing regions both exhibit a similar up-then-down pattern. But death rates in the growing region flattened in 1994 while the decrease in the non-growing region accelerated in 1995. Violent death rates increased in both regions after 1995, but the average rate of increase in the growing region became much steeper than in the non-growing region. In contrast with this parallel-then-divergent pattern in violent death rates, death rates from disease fell somewhat more steeply in the growing than in the non-growing region from 1990 through 1998, when there was a sharp upturn in the growing region. The evolution of death rates from disease can be seen in Figure 3b. The evolution of death rates from disease can be seen in Figure 3b.

Figures 4a and 4b show a similar picture in the context of a 3-region analysis that separates the DMZ from other growing departments. Here too, violent death rates in the non-DMZ growing region flatten in 1994 while death rates in the non-growing region turn more sharply downwards in 1995, so that by 1995, the gap in death rates between the two region types had closed. The main difference between this picture and that in Figure 3a is that violent death rates in the DMZ fell through 1994, after which they

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The non-growing region omits Antioquia, Valle de Cauca and Bogota, the departments with Colombia's three largest cities. Deaths rates were coded from vital statistics micro data obtained from the Colombian statistical agency, DANE. Violent deaths are defined here as homicides, suicides, deaths from military and insurgent activity (not a distinct category in all years), and a small number of non-accident deaths by external causes not elsewhere classified. Over 90% of violent deaths are homicides. For additional details see the Data Appendix.
Competing risks complicate the interpretation of the decline in disease death rates since some of those who die by

²¹ Competing risks complicate the interpretation of the decline in disease death rates since some of those who die by violence may have otherwise died of disease. Still, it seems likely that an environment of deteriorating public health would turn up in higher disease death rates (a pattern observed in the DMZ after the government ceded control). The competing risks problem is likely mitigated by the fact that those most likely to die from disease (the very old and very young) are least likely to die by violence. As a check on this problem, we looked at infant and child mortality by region type. These data also show relative improvement in the growing region until 1998.

increased sharply until a 1998 peak. In contrast with the relative increase in violent death rates in the growing region, death rates from disease improved steadily relative to the non-growing region beginning in 1992. Although somewhat more volatile, death rates from disease in the DMZ also roughly tracked those in the non-growing region until 1998 when the non-growing region saw a relative improvement.

A possible complication in the analysis of death rates is the quality of the population statistics used for the denominator. We used census-based 5-year estimates and population projections published by DANE (1998) for 1990, 1995, and 2000, linearly interpolating statistics for the in-between years. As noted above, however, the 1990s were marked by considerable population movement so the population denominator may be inaccurate.²² An alternative strategy which avoids this problem is to look at violent death rates relative to death rates from other causes. After transformation to log odds, this approach can be motivated by a multinomial logit model for the risk of death by cause.

To describe the logit strategy more formally, let v_{jt} denote the number of violent deaths in department j and year t and let n_{jt} denote the number of deaths from all other causes. Let p_{jt} denote the corresponding population statistics. Write the probability of violent and non-violent death as

$$v_{it}/p_{it} \equiv exp(\alpha_{it}(v)) / [1 + exp(\alpha_{it}(v)) + exp(\alpha_{it}(n))],$$

$$n_{it}/p_{it} \equiv exp(\alpha_{it}(n)) / [1 + exp(\alpha_{it}(v)) + exp(\alpha_{it}(n))],$$

where $\alpha_{jt}(v)$ and $\alpha_{jt}(n)$ are functions of region and year. We assume further that $\alpha_{jt}(v)$ can be modeled as an additive function of region (B_j) and year (Δ_t) main effects plus growing-region/year interaction terms induced by the shift of coca production to Colombia, while $\alpha_{jt}(n)$ has additive region and year main effects only. In this framework, non-zero estimates of the interaction terms B_{0s} and B_{1s} in the grouped-logit equation

$$ln(v_{it}/n_{it}) \equiv \alpha_{it}(v) - \alpha_{it}(n) \equiv B_i + \Delta_t + \Sigma_s B_{0s} g_{is} + \Sigma_s B_{1s} d_{is} + v_{it}$$

provide evidence of region-specific shocks that increased the risk of violent death.

This strategy is illustrated in Figure 5, which plots the residual from a regression of $ln(v_{jt}/n_{jt})$ on region effects (i.e., deviations from group means) using the 14-department classification scheme,

²² These statistics differ from the population figures in Table 2, which are based in the 1993 census.

separating the DMZ from other growing departments as in Figure 4.²³ The logit plot shows the same initial pattern of up-then-down log-odds of violent death in both growing and non-growing regions, with a more stable then increasing pattern of log-odds of violent death in the growing region after 1993 and especially after 1995, while the odds of death by violence were decreasing in the non-growing region from 1993-95. This is broadly similar to the pattern exhibited by log death rates in Figure 4a. The log-odds of violent death in the DMZ also turned sharply upwards in 1995.

On balance, the figures suggest that beginning sometime between 1994 and 1996, violent death rates in the non-DMZ growing region became markedly higher than what should have been expected based on pre-1994 trends. The pattern was generally similar in the DMZ, where the FARC presence was stronger. Moreover, the increase in violent death rates contrasts with a gradually improving disease environment in this period, both nationwide, and in the growing relative to the non-growing region. The contrast in trends for violence and disease mortality weighs against the notion of a secular deterioration in infrastructure or social systems that caused the increase in violence.

Regression Estimates

To quantify the relative increase in violent death rates in the growing region, growing-region/year interaction terms were estimated using the following equation:

$$ln(v_{ajt}/p_{ajt}) = \mu_a + \beta_j + \delta_t + \Sigma_s \alpha_{0s} g_{js} + \Sigma_s \alpha_{1s} d_{js} + \varepsilon_{ajt}.$$
(3)

The dependent variable, $ln(v_{ajt}/p_{ajt})$, is the log death rate in cells defined by 10-year age groups (indexed by a), department (indexed by j), and year (indexed by t). The term μ_a is an age effect, while other parameters are defined as in equation (1). Also, as with the analysis in the previous section, some models include trends for each department type.²⁴

Unweighted estimates of α_{0s} and α_{1s} are reported in Panel A of Table 7, while population-

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²³ Year effects are left in the plotted series so the common trend is visible.

²⁴ For purposes of estimation, the sample was expanded slightly to include ages 15-64 to accommodate the 10-year age groups. Data are analyzed for age-specific cells to control for changes in the age distribution due to migration and because mortality trends tend to be age-specific. Standard errors were adjusted for department clustering, though as far as asymptotic inference based on cell size goes, these are population regression coefficients.

weighted estimates are shown in Panel B. The table reports results with and without medium producers. The unweighted estimates of α_{0s} in Column (1) show an insignificant 12.3 percent (log point) increase in mortality in 1994 in the non-DMZ growing region, with no effect in 1993 (a specification check). The unweighted growing-region effect increases through 1997, while the DMZ effect increases through 1998. Few of the individual yearly effects are significant, but, as we show in later tables the consistent pattern across regions and over times generates significant pooled estimates. Weighting also tends to increase precision, though it leads to somewhat smaller effects in models without trends. Omission of medium producers leads to larger and more significant effects.²⁵

Inclusion of region-specific trends leads to less precise estimates, though the weighted estimates are markedly larger when estimated with trends than without. None of the estimates using the 14-department classification are significant when estimated with trends. On the other hand, weighted models with trends generate significant effects as early as 1994 when estimated without medium producers. These estimates are reported in Column (7). Although some of the estimates are quite large in percentage terms, it should be noted that the number of deaths involved is often small, particularly in sparsely populated rural departments.

Appendix Table A2 reports the results of robustness checks for the basic mortality estimates similar to the robustness checks reported in Panels A-C of Table A1 for the rural results. Again we focus on models combining growing and DMZ departments, retaining the yearly-interaction format of Table 7. The variations reported here include a benchmark with no modifications in Columns (1) and (5) (other than the pooling of both department types), the addition of time-varying department-average rainfall, and the addition of GSP trends. These modifications leave the results unchanged. A more detailed exploration of results involving differential trends and treatment effects for departments with a previous guerilla presence is discussed at the end of this section.

²⁵ DMZ effects are (by construction) identical with and without medium producers since the medium producers are a subset of the non-DMZ growing region.

Stacked Urban and Rural Mortality Estimates

The link between coca production and increased death rates may run through a number of channels. Like other illegal industries, cocaine is associated with violence and intimidation. Moreover, as we noted at the outset, coca is widely believed to provide revenue for guerilla and paramilitary groups in rural areas, either through taxation, protection rackets, or direct control of production. The resulting revenue makes it easier for violent groups to obtain weapons and recruits, and generally step up their activities, which are largely in rural areas. In contrast, most of the violence associated with the cocaine *trade* occurred in large cities (most dramatically, in Medellin). Violence that arises through increased coca *cultivation* should therefore be a bigger problem in the countryside than in cities. To substantiate this, we estimated growing region effects on violent death rates separately for urban and rural victims.²⁶

For the most part, estimates by urban/rural status indeed point to a stronger link between coca penetration and violent death rates in rural than urban areas. This is documented in Table 8, which reports estimates of equation (3) for rural residents in Panel A and for urban residents in Panel B.²⁷ For example, the rural estimates of α_{0s} in Column (1) are 0.31 (s.e.=0.15) for 1995, 0.36 (s.e.=0.19) for 1996, and 0.4 (s.e.=0.221) for 1997. The corresponding estimates for urban deaths, reported in Column (1), Panel B, are 0.05, 0.17, and 0.11, none of which are significant. Overall, the contrast between Columns (1)-(2) in Panels A and B shows much larger effects in rural than urban areas. This pattern also persists in models that include trends (results reported in Columns (3) and (4)), with some of the effects on rural areas large and significant (and the corresponding trends large and negative). The contrast in results by urban/rural status using a sample that omits medium producers, reported in Columns (5)-(8), generally also shows marked larger effects in rural areas, with the exception of some entries in Column (5).

²⁶ DANE mortality files identify the type of area in which the deceased lived and the location of death. We defined urban/rural status by type of residence since hospitals where victims may die are mostly found in cities. For the purposes of our analysis, the deceased was identified as urban when residence was coded as "cabecera municipal." The urban residence variable is available only from 1992.

The urban residence variable is available only from 1992.

The urban/rural distinction is used for the numerator but ignored in the population denominator. Since the model is in logs, this probably provides a reasonable approximation to an analysis of true death rates by urban/rural status. Estimates in Table 8 are unweighted since we do not have departmental population estimates by urban/rural status for intercensal years. An unweighted analysis may be more appropriate in any case, since the regressor of primary interest varies at the department-year level.

Estimates by Age and Previous Guerilla presence

The possibility of differential effects by demographic subgroup and across departments motivates two further analyses, one by age and one allowing for a distinction between departments with and without a strong pre-existing guerilla presence. Violence caused by the criminal activities of those involved in the cocaine industry may be concentrated among young men. Consistent with this, we found no evidence of an effect of coca expansion on violent deaths among women. Civilian victims of the Colombian civil conflict, on the other hand, seem to come from all age groups, though they are also mostly male (see, e.g., the accounts in Kirk, 2003). We therefore estimated a version of the models used to construct the estimates in Table 8, pooling interaction terms across growing regions and years as in Table 6, but allowing effects to differ for men under and over 35. These estimates, reported in Table 9, show essentially similar impacts on younger and older men. This supports the view that the accelerating violence in growing regions is probably more conflict-related than criminal in nature.

Models allowing for separate trends and separate treatment effects in areas with a pre-existing guerilla presence were estimated both as a robustness check and to explore the variability in treatment effects across departments. This is motivated by the fact that a strong pre-existing guerrilla presence is prima facie evidence of an environment hospitable to insurgents. The baseline strength of the guerilla presence is measured in two ways, first using the number of FARC attacks in 1994, and second, summing attacks by three guerilla groups plus the AUC (a paramilitary group) in 1995. In particular, we coded two dummies for departments with guerilla activity above the median for these measures. The estimates come from a pooled specification similar to that used to construct the estimates in Table 6. In this case, however, in addition to the treatment effect interactions, the models allow for 1994, 1995-97 and 1998-2000 interactions with guerilla activity dummies to control for trends specific to (both growing and nongrowing) departments with a strong pre-existing guerilla presence.

²⁸ In principle, an earlier measure of guerilla activity might be preferable, but data for the earlier period are spotty. We note, however, that there appears to be considerable overlap between the 1994 and 1995 measures used here and the departments listed as guerilla strongholds in an earlier study by the Federal Research Division (1988) of the Library of Congress.

The relationship between coca growth and accelerating violence is stronger for departments with substantial early guerilla activity. This is documented in Table 10, which reports estimates from models using both measures of guerilla activity. For example, when growing regions are defined using the 14-department definition, estimates using the first measure show no effect of coca growth on violence in departments without much early activity. Models estimated without medium producers generate effects in both types of departments, but the effects are consistently larger in departments that had substantial earlier guerilla activity. These findings are therefore consistent with the view that an initial guerilla presence facilitated the increase in insurgent activities due to coca.

VI. Summary and Conclusions

The disruption of the Andean air bridge provides an opportunity to assess the impact of coca production on Colombia's rural population and to study the link between a resource boom and violence. On the economic side, we find some evidence of increases in self-employment income, though not in the likelihood of having income from this source, in the probability of working more generally, or in wage and salary earnings. The increase in self-employment income is estimated to be on the order of 13-29 log points, a substantial gain. There is also some evidence of an increase in boys' labor supply.

Because the gains to increased coca cultivation appear to be fairly concentrated, it seems unlikely that increased coca production raised overall standards of living in growing areas. The absence of wider gains may be due to the fact that coca has few links with other sectors or to extortion on the part of insurgents and paramilitary forces and the fact that coca finances a conflict that reduces economic activity outside of the coca sector.²⁹ Consistent with this latter story, our results show increased violent death rates in growing areas after the increase in coca cultivation, though it should be noted that these results are weaker in models that include department trends. On the other hand, the findings are reasonably consistent in models with time-varying controls for rainfall and various other trend controls.

We cannot conclusively identify the channels through which coca cultivation might abet violence,

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²⁹ Pshisva and Suarez (2004) suggest that the risk of kidnapping reduces investment by Colombian firms.

but differences in effects by urban/rural status are consistent with the notion that coca supports rural insurgents and paramilitaries. The fact that older and younger men are similarly affected also points to the civil conflict, as opposed to criminal activity, as a primary cause of increased rural violence as does the finding that violence increased more in the rural parts of growing departments with substantial previous conflict. Remarkably, the increase in rural violence occurred against a backdrop of generally improving public health as measured by death rates from disease.

Our results provide an interesting case study of a situation where increases in income did not lead to a reduction in civil conflict, but rather fueled the fires of unrest. This contrasts with the more optimistic picture in Miguel, Satyanath, and Sergenti (2004), but is in line with journalistic accounts of the role played by blood diamonds in Africa's civil wars and economic theories of insurrection as extraction or extortion (e.g., Grossman, 1991; Collier and Hoeffler, 2004). Coca may indeed be emblematic of a particular sort of resource curse associated with goods for which there is a well-developed black market.

DATA APPENDIX

Colombian Rural Household Surveys

The analysis here uses the "Encuesta Rural de Hogares", the rural component of the "Encuesta Nacional de Hogares," which became "Encuesta Continua de Hogares" (a panel) in 2001. The rural household survey was first conducted as a pilot in 1988. The survey was conducted again in December 1991 after the sampling methodology was updated and then used on a consistent basis every September starting in 1992 until 2000. The survey collects data on a representative sample from 23 departments in four rural regions: Atlantic Region (which includes the departments of Atlantico, Cordoba, Magdalena, Sucre, Cesar, La Guajira, and Bolivar); Pacific Region (which includes the departments of Choco, Nariño, Cauca, and Valle del Cauca); the Central Region (which includes the departments of Antioquia, Caldas, Huila, Tolima, Quindio, Risaralda, and Caqueta); and the Eastern Region (which includes the departments of Norte de Santander, Santander, Boyaca, Cundinamarca, and Meta).

Rural definition

The survey uses the following criteria to identify the rural population. The rural population includes:

- 1. The population of the city where the county's government is located if the city has less than 10,000 inhabitants.
- 2. The population of the city where the county's government is located if the city has more than 10,000 inhabitants and it meets one of the following characteristics:
 - (a) the percentage of residents in the city does not exceed 50% of the population in the entire county,
 - (b) the percentage of the active population engaged in agricultural activities exceeds 50%, or
 - (c) the percentage of housing units without basic services (water, electricity, etc.) exceeds 20%.
- 3. All population living in towns with less than 10,000 inhabitants.
- 4. All population not living either in cities or towns.

Sampling methodology

The sample for the survey is taken from the universe of the Census population living in private households. The sampling methodology consists of first generating strata according to geographical location and socioeconomic level; then, randomly drawing 'municipios' (the equivalent of counties in the U.S.) from these strata; next, randomly drawing neighborhoods from these 'municipios'; and, finally, randomly drawing blocks and then households from these neighborhoods. To facilitate the collection of information, households are grouped into 'segments' of 10 households on average. The typical year includes approximately 8,500 households, but the sample has increased over time. In particular, the sample size increased in 1996. The survey collected data from 148 municipios in 1992-1995, but it collected data from 197 municipios in 1996-2000.

In addition, the survey methodology changed as follows in 1996. First, between 1992 and 1995 the sample was drawn from the 1985 Census, while starting in 1996 and until 2000 the sample was drawn from the 1993 Census. Second, starting in 1996, interviewers were required to revisit households, which generated an increase in response rates.

Sample weights

The survey weights include factors of adjustment to account for changes in subsampling and for non-response. So, we use the weighted data in our analysis to take account of the 1996 changes. In particular, the weights are estimated as:

$$W = (1/P) \times S \times (I_S/N_S)$$

where P is the probability of an individual being sampled and S is a weight given to segments. S equals 1 unless the number of households within the segment exceeds 10. The last term is the ratio of the number of households actually interviewed within a segment, I_S , and the number of households selected for interviewing within a segment, N_S , so it captures the response rate within a segment.

Since the average number of children per household is around 3, we generate within household averages for the children's data in order to avoid multiple observations per household. Likewise, since the weights are individual weights, we construct household weights by summing up the individual weights for all children within the household.

Top-coding and imputation

Labor market information is collected from individuals aged 10 and up. We impute zeros for the employment and hours of 8 and 9 year olds in the descriptive statistics in Table 5. Hours are collected from all employed workers, including salaried and wage workers as well as self-employed workers. Wage and salary earnings were collected for all jobs in 1992-1999. In 2000, wage and salary earnings were collected separately for the main job and for secondary jobs, so we exclude 2000 from the wage and salary regressions. Yearly self-employment income is collected separately as earnings from business and commercial activities and family-based agricultural production. In the original data, earnings and self-employed income were top coded only between 1992 and 1995. We impose uniform top-coding by applying a cap at the 95th percentile (including zeros) for each year. In addition, we imputed the mean earnings and self-employment income by department and year for all those individuals who reported having earnings or self-employed income but did not report an amount.

Colombian Urban Household Surveys

We also use the urban component of the "Encuesta Nacional de Hogares," from 1992-2000. Coverage for the urban component of the survey is more limited than for the rural survey. In particular, the urban survey excluded the departments of Magdalena, Caqueta and Choco in 1992-1995. For the urban/rural analysis we used the 20 departments which are covered in both surveys in 1992-2000. Also, because Caqueta, one of the two DMZ departments, is not in the urban survey during the initial years, the urban/rural analysis pools Meta, the other DMZ department, with the non-DMZ growing departments, so that these specifications only include non-DMZ growing interaction terms.

The sampling methodology and sample weights used in the urban component are similar to those used in the rural component. The urban population includes all people not included under the rural definition.

As in the rural component of the survey, wage and salary earnings were collected for all jobs in 1992-1999, but in 2000 wage and salary earnings were collected separately for the main job and for secondary jobs, so the 2000 data are excluded from the wage and salary regressions. Yearly self-employment income is collected separately as earnings from business and commercial activities and the sale of domestically produced goods. In the urban component of the survey, earnings and self-employed income

were top coded only between 1992 and 1996, so we apply the same uniform top-coding as used in the rural survey by applying a cap at the 95th percentile (including zeros) for each year. We also imputed mean earnings and self-employment income by department and year for all those individuals who reported having earnings or self-employed income but did not report an amount.

Mortality Detail Files

We obtained mortality detail files from the Colombian national statistical agency, DANE, for 1990-2001. These files, the source of published vital statistics (e.g., http://www.dane.gov.co/inf_est/vitales.htm), show individual death records, with basic demographic information on the deceased and cause of death. The 1990 and 1991 files did not include reliable urban/rural codes and are therefore omitted from the sample used to construct Table 8. The 2001 file also had some inconsistencies (the file was provisional) and was therefore dropped.

Cause of death

We aggregate detailed causes of death on a consistent basis from year to year into the following larger groups: homicide and suicide, accident and other non-violent trauma, disease, other causes, and other violent deaths. The violent death rate used here is the sum of homicide and suicide plus other violent deaths. Data after 1997 show separate categories for general external causes not identified as accidents and deaths due to actions by state and guerilla forces. These two categories appear to correspond to the "other violence category" from previous years.

Location information

Our construction of death rates by department and year is for department of death and not residence. Urban/rural status, however is by area of residence. This is coded somewhat differently from year to year. We established a consistent urban/rural by coding as urban those listed as living in "cabecera municipal" and coding the institutionalized as non-urban. Those with missing urban/rural status (about 1/16 of deaths) are omitted from the analysis used to produce Table 8.

Match to population information

As noted in the text, population statistics for each department-year-age(5 year)-sex category were obtained from DANE (1998) for 1990, 1995, and 2000. The Colombian census used for these data was conducted in 1993, so for other years we use inter-censal estimates and projections. We interpolated using 5-year growths for each cell.

Finally, we aggregated mortality counts to match 5-year age bands, and then matched to the relevant population denominators.

Miscellaneous

Farm hectares used to calculate percent of cultivated land devoted coca in the text discussion come from various published and web sources distributed by the Colombian statistical agency (DANE) and the Ministry of Agriculture. Details are available on request.

Data on GSP are from the DANE web site. We divided aggregate GSP by the 1993 population figures in Table 2 to get per-capita GSP.

Data on 1994 FARC and 1995 aggregate guerilla activity come from police records and was kindly provided to us by Gustavo Suarez. These data measure the number of attacks/actions taken by each of these groups, including terrorist activities, extortion, attacks on rural and urban populations, harassment, attacks on infrastructure, attacks on airplanes, arms smuggling, and confrontations with other groups. The 1995 data sum actions by the FARC, ELN, and EPL guerrilla groups, plus the paramilitary AUC. We coded dummies for departments above the median of each (8 for FARC in 1994, 22 for total in 1995). Half of the 14 growing departments had substantial previous guerilla activity using the 1994 measure; 8 of 14 growing departments had substantial previous guerilla activity using the 1995 measure.

Data on rainfall were purchased from the Colombian Institute of Hydrology, Meteorology, and Environmental Studies (IDEAM). These data measure yearly precipitation measured in centiliters as well as the total number of days of precipitation during the year. We have information for all departments for the station corresponding to the main airport in each department. The data include complete yearly data for the period from 1986 to 2005, thus covering the entire period for which we have household and homicide data.

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Fig. 1. Production of Coca Leaf in Colombia, Peru and Bolivia 1990-2000

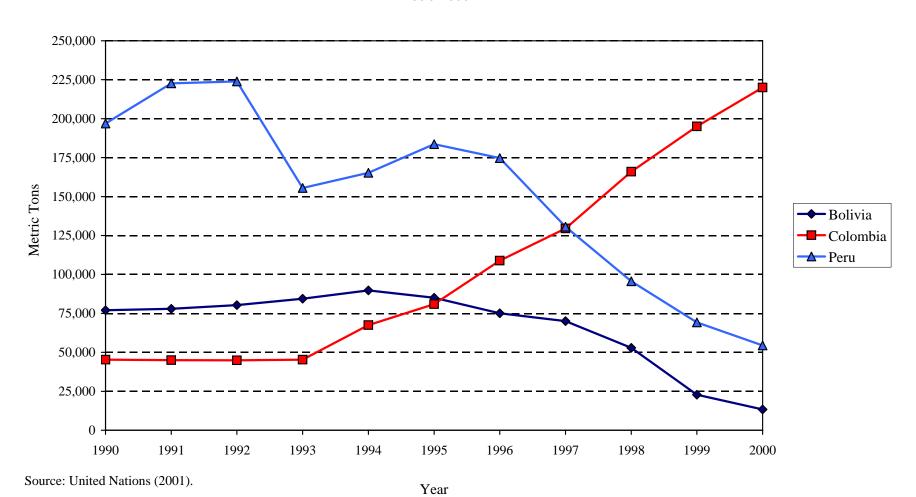
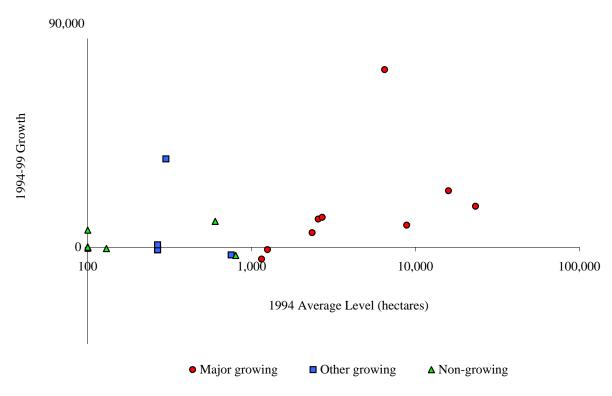
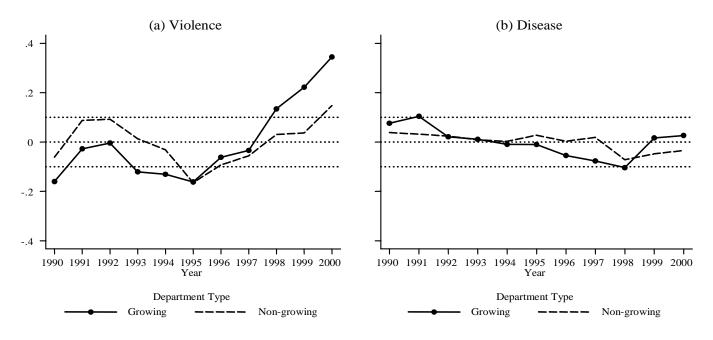


Fig. 2. Coca Cultivation: 1994-99 growth as a function of 1994 levels



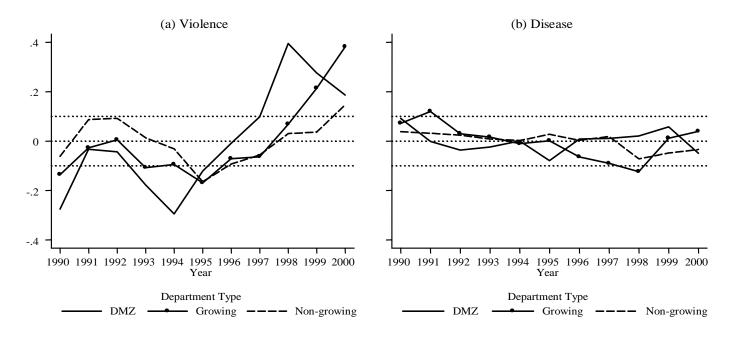
Notes: Scales are logarithmic. The 100 hectare base group includes 100 or less.

Figure 3. Death Rates for Men Aged 15 - 59.



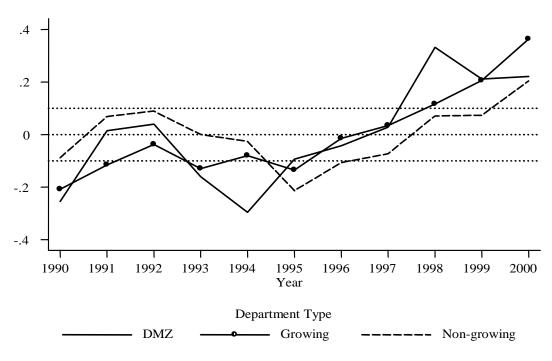
- Notes:
 1. Log rates, relative to average by department type.
 2. Non-growing omits Antioquia, Valle, and Bogota DC.

Fig. 4. Death Rates for Men Aged 15 - 59



- Notes:
 1. Log rates, relative to average by department type.
 2. Non-growing omits Antioquia, Valle, and Bogota DC.

Fig. 5. Death Rates – Logit(Violence/Total) -- for Men Aged 15 - 59



Notes:
1. Logits, relative to average by department type.
2. Non-growing omits Antioquia, Valle, and Bogota DC.

TABLE 1. FIRST STAGE FOR COCA CULTIVATION GROWTH

	IADL	L 1.11N3		994 to 199	9	71101V OF		994 to 200	00
Treatment group	Parameter	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
			A. Wi	th DMZ de	epartments				
	_								
9 growing dept.	Intercept	735	207			506	292		
	Growing	(1,749)	(2,112)			(2,024) 8,748	(2,458)		
	Glowing	7,554 (3,445)	8,082 (3,724)			(3,876)	8,961 (4,358)		
14 growing dept.	Intercent	(3,443)	(3,724)	207		(3,870)	(4,336)	292	
11 growing dept.	тистесри			(2,053)				(2,348)	
	Growing			6,100				6,127	
	C			(3,152)				(3,604)	
Linear	Intercept				1,658				2,147
					(1,708)				(1,989)
	Hectares				.553				.362
					(.321)				(.373)
Includes medium		4 1		441		1		441	
producers?		control	no	treated	yes	control	no	treated	yes
R^2		0.134	0.121	0.108	0.087	0.141	0.107	0.085	0.029
Λ		0.154	0.121	0.100	0.007	0.141	0.107	0.005	0.02)
			B. With	out DMZ	departments				
9 growing dept.	Intercept	735	207			506	292		
y growing dept.	тистесри	(1,845)	(4,198)			(2074)	(4,911)		
	Growing	8,434	8,966			9,533	9,746		
	C	(3,883)	(4,198)			(4,364)	(4,911)		
14 growing dept.	Intercept	, , ,		207		, ,	, , ,	292	
				(2,115)				(2,414)	
	Growing			6,287				6,112	
				(3,400)				(3,879)	
Linear	Intercept				1,699				2,201
					(1,759)				(2,057)
	Hectares				.697				.328
					(.398)				(.466)
Includes medium producers?		control	no	treated	yes	control	no	treated	yes
R^2		0.140	0.125	0.106	0.091	0.141	0.105	0.079	0.017

Notes: The table reports estimates of the change in cocaine cultivation on 1994 levels for the 33 Colombia Departments (states). The 1994 variable is the average of 4 measures from Thoumi et al. (1997). The 1999 and 2000 data are police estimates, reported in Government of Colombia (2002).

TABLE 2. DESCRIPTIVE STATISTICS BY REGION TYPE

				Enre	ollment		
Region type	Department	Population	% Urban	% Primary		Homicide Rate	Per capita
8	_ 	1993	1993	1995	1995	1990-1993	GSP 1993
		(1)	(2)	(3)	(4)	(5)	(6)
Non-growing	Santafe de Bogota, DC	4,945,448	99.7	60.8	75.2	178.8	3,090
11011-growing	Amazonas	37,764	50.4	56.3	33.2	35.3	1,744
	Antioquia	4,342,347	72.0	75.1	59.1	718.7	2,358
	Arauca	137,193	63.6	90.0	38.8	226.3	6,703
	Atlantico	1,667,500	93.7	68.2	67.9	65.2	1,633
	Boyaca	1,174,031	42.5	65.8	47.4	124.1	1,746
	Caldas	925,358	64.7	66.2	56.3	251.6	1,552
	Casanare	158,149	54.7	77.9	34.5	153.1	4,311
	Choco	338,160	38.5	66.6	32.0	65.6	845
	Cordoba	1,088,087	48.2	89.5	50.5	66.7	1,129
	Cundinamarca	1,658,698	54.7	71.7	55.3	126.4	2,112
	Huila	758,013	60.0	73.9	47.7	132.6	1,663
	Quindio	435,018	83.8	64.7	65.0	173.0	1,556
	Risaralda	744,974	81.3	66.1	59.1	303.4	1,633
	San Andres y Providencia	50,094	70.4	69.7	84.2	44.1	3,585
	Santander	1,598,688	68.9	68.9	53.3	192.7	2,111
	Sucre	624,463	67.1	96.6	51.3	38.5	892
	Tolima	1,150,080	60.7	71.9	55.5	140.9	1,626
	Valle del Cauca	3,333,150	85.3	71.9	62.6	311.5	2,382
	All 19 non-growing	25,167,215	75.5	70.9	59.8	271.6	2,205
	Non-growing	12,546,270	64.5	72.7	54.0	140.7	1,756
	(w/o Bogota, Antioquia, and Val	le de Cauca)					ŕ
9-Dept	Bolivar	1,439,291	68.6	75.4	50.4	41.1	1,628
growing	Cauca	979,231	36.7	73.8	36.3	170.0	1,065
	Guaviare	57,884	36.9	59.3	17.4	130.5	2,955
	Narino	1,274,708	42.9	62.4	33.7	58.7	895
	Putumayo	204,309	34.6	75.4	28.0	170.7	1,042
	Vaupes	18,235	24.8	74.9	21.4	2.8	1,971
	Vichada	36,336	24.1	54.2	16.6	39.3	1,640
	All 9-dept w/o DMZ	4,009,994	49.8	70.6	39.4	86.5	1,249
9-Dept, DMZ	Caqueta	311,464	46.0	76.4	31.7	205.2	1,301
	Meta	561,121	64.0	72.9	52.2	204.7	2,190
	All DMZ	872,585	57.5	74.3	44.3	204.9	1,872
Medium	Cesar	729,634	62.9	82.1	50.7	146.2	1,354
producers	Guainia	13,491	30.4	43.3	17.1	45.3	1,871
	La Guajira	387,773	64.3	72.5	58.5	135.8	2,020
	Magdalena	882,571	64.0	67.7	41.0	70.8	1,272
	Norte de Santander	1,046,577	70.8	67.8	42.4	241.2	1,169
	All medium producers	3,060,046	66.0	71.7	45.6	150.6	1,354
	All Departments	33,109,840	71.0	71.1	55.2	235.8	2,002

Source: Colombia Estadistica, 1993-1997 and tabulations of vital statistics. Columns (2), (3) - from table 2.1.2. Column (3): % of pop. living in cabecera. Columns (4), (5) - from tables 10.2.1 and 10.3.1. Column (4): primary enrollment divided by pop. aged 5-9 plus 60% of the pop. aged 10-14. Column (5): secondary enrollment divided by 40% of the pop. aged 10-14 plus pop. aged 15-19. Homicide rates are for men aged 15-59, per 100,000. Column (6): 1993 per capita GSP in US dollars.

TABLE 3. DESCRIPTIVE STATISTICS FOR RURAL SURVEY

		workers 1 women)		Workers nen)	Во	oys	Gi	irls	_	Workers
Variable	1992 (1)	1997 (2)	1992 (3)	1997 (4)	1992 (5)	1997 (6)	1992 (7)	1997 (8)	1992 (9)	1997 (10)
Employed	0.658 (0.474)	0.647 (0.478)	0.950 (0.219)	0.931 (0.253)	0.360 (0.436)	0.283 (0.413)	0.095 (0.264)	0.077 (0.250)	0.600 (0.450)	0.506 (0.458)
Hours Worked per month	142.3 (117.3)	131.8 (115.0)	219.3 (76.8)	200.3 (85.2)	65.6 (86.9)	44.9 (74.3)	15.4 (47.8)	11.2 (39.7)	121.2 (103.0)	94.1 (96.6)
Monthly Wages	74,098 (115,512)	81,461 (127,287)	115,439 (126,002)	123,636 (137,069)					61,038 (78,522)	58,921 (84,989)
Positive Wages	0.362	0.369	0.551	0.555					0.418	0.350
SE Income (5% top code)	337,712 (815,718)	352,969 (865,459)	557,381 (1,026,941)	551,260 (1,048,327)					61,196 (311,898)	65,253 (291,899)
Positive SE Income	0.247	0.259	0.348	0.371					0.066	0.080
Enrolled	0.017	0.028	0.014	0.024	0.694	0.779	0.757	0.815	0.397	0.486
Age	36.43 (10.65)	37.02 (10.62)	36.60 (10.68)	37.12 (10.63)	11.91 (2.10)	11.94 (2.18)	11.87 (2.15)	11.90 (2.19)	16.16 (1.95)	16.22 (1.90)
HH Size	5.60 (2.56)	5.32 (2.49)	5.56 (2.60)	5.31 (2.51)	6.60 (2.36)	6.35 (2.41)	6.66 (2.36)	6.41 (2.43)	6.57 (2.48)	6.35 (2.57)
Migrant	0.282	0.316	0.280	0.312	0.133	0.161	0.163	0.160	0.161	0.182
Single	0.229	0.221	0.284	0.277	1.00	0.999	0.990	0.981	0.981	0.970
Male	0.496	0.597								
Growing (14-dept)	0.235	0.303	0.234	0.309	0.262	0.332	0.250	0.336	0.238	0.332
Growing (9-depts)	0.137	0.181	0.133	0.180	0.147	0.191	0.138	0.198	0.138	0.197
DMZ	0.014	0.039	0.014	0.039	0.013	0.040	0.014	0.044	0.011	0.034
Max N	13,550	19,184	6,641	9,801	2,602	3,513	2,477	3,253	2,040	2,881

Notes: Adult workers include men aged 21-59. Boys and girls are aged 8-16. Labor market outcomes for boys and girls are only reported for those over 10 years of age. Teenage workers are boys aged 13-20. Wages and self-employment income include zeros and are in real (1998) pesos, about 1,400 to the US dollar.

TABLE 4a. ADULT LABOR MARKET OUTCOMES - YEARLY INTERACTIONS, POOLED GROWING & DMZ

	Male and Fen	nale Workers	<u> </u>			IVI	en Only		
Positive	e SE Income	Log S	SE Income	En	nployed	Log Ho	urs (All Jobs)	_	thly Wage and y Earnings
All	w/o Migrants	All	w/o Migrants	All	w/o Migrants	All	w/o Migrants	All	w/o Migrants
(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
-0.034		-0.025		-0.014	-0.019			0.014	0.027
(0.025)	(0.029)	(0.161)	(0.137)	(0.013)	(0.016)	(0.024)	(0.026)	(0.029)	(0.036)
0.009	0.002	0.159	0.190	-0.002	-0.008	0.004	0.006	-0.011	-0.008
									(0.030)
(0.017)	(0.020)	(0.1.0)	(0.1.0)	(0.012)	(0.010)	(0.021)	(0.020)	(0.020)	(0.020)
0.042	0.048	0.362	0.414	0.003	-0.010	0.034	0.048	-0.008	0.002
(0.039)	(0.035)	(0.155)	(0.171)	(0.017)	(0.020)	(0.031)	(0.037)	(0.055)	(0.068)
0.025	0.034	0.270	0.292	-0.005	-0.013	-0.001	0.002	0.010	0.028
(0.027)	(0.026)	(0.124)	(0.115)	(0.015)	(0.146)	(0.035)	(0.042)	(0.053)	(0.059)
-0.020	-0.009	0.302	0.291	0.0004	-0.012	0.053	0.035	0.103	0.107
(0.033)	(0.029)	(0.154)				(0.023)		(0.067)	(0.070)
()	(3.2.2)	(/	(3.7.3.7)	()	(3.2.2)	()	(3.13.2)	(,	(******)
0.007	0.016	0.205	0.150	0.016	0.007	0.006	0.005	0.069	0.067
(0.023)	(0.017)	(0.122)	(0.132)	(0.014)	(0.017)	(0.023)	(0.023)	(0.048)	(0.046)
0.042	0.051	0.255	0.220	0.002	0.012	0.065	0.005		
(0.033)	(0.029)	(0.158)	(0.1/4)	(0.014)	(0.015)	(0.024)	(0.025)		
148,306	100.638	40.338	28.912	74.781	50.914	69.144	46,770	34,451	22,257
_	All (1) -0.034 (0.025) 0.009 (0.017) 0.042 (0.039) 0.025 (0.027) -0.020 (0.033) 0.007	(1) (2) -0.034	All (1) w/o Migrants (2) All (3) -0.034 (0.025) -0.045 (0.029) -0.025 (0.161) 0.009 (0.002 (0.159) (0.159) (0.017) (0.020) (0.146) 0.042 (0.039) (0.035) (0.155) 0.025 (0.034 (0.027) (0.026) (0.124) -0.020 (0.027) (0.026) (0.124) -0.020 (0.033) (0.029) (0.154) (0.025) (0.154) 0.007 (0.016 (0.023) (0.017) (0.122) (0.043 (0.051) (0.255) (0.033) (0.029) (0.158)	All (1) w/o Migrants (2) All (3) w/o Migrants (4) -0.034 (0.025) -0.045 (0.029) -0.025 (0.161) 0.080 (0.137) 0.009 (0.002 (0.159) 0.190 (0.146) 0.190 (0.145) 0.042 (0.048 (0.039) 0.362 (0.144) 0.414 (0.039) 0.025 (0.034 (0.027) 0.270 (0.292 (0.171) 0.025 (0.027) 0.026) (0.124) (0.115) -0.020 (0.033) -0.009 (0.154) 0.291 (0.189) 0.007 (0.016 (0.023) (0.017) 0.205 (0.150 (0.132) 0.043 (0.031) (0.029) 0.255 (0.230 (0.174) 0.033) (0.029) (0.158) (0.174)	All (1) w/o Migrants (2) All (3) w/o Migrants (4) All (5) -0.034 (0.025) -0.045 (0.029) -0.025 (0.080) -0.014 (0.0137) -0.014 (0.0137) 0.009 (0.002) 0.159 (0.190) -0.002 (0.013) 0.042 (0.048) 0.362 (0.144) 0.003 (0.039) 0.035) (0.155) (0.171) (0.017) 0.025 (0.027) 0.026) (0.124) (0.115) (0.015) -0.020 (0.026) 0.302 (0.124) 0.291 (0.004) 0.006) 0.007 (0.016) 0.205 (0.154) 0.189) (0.016) 0.007 (0.016) 0.205 (0.122) 0.150 (0.014) 0.043 (0.023) 0.051 (0.025) 0.230 (0.0132) -0.003 (0.014) 0.043 (0.033) (0.029) (0.158) (0.174) (0.014)	All (1) w/o Migrants (2) All (3) w/o Migrants (4) All (5) w/o Migrants (6) -0.034 (0.025) -0.045 (0.029) -0.025 (0.161) 0.080 (0.137) -0.014 (0.019) -0.019 (0.013) (0.016) 0.009 (0.002 (0.159) 0.159 (0.146) 0.190 (0.013) -0.002 (0.016) -0.008 (0.017) (0.020) (0.146) (0.145) (0.013) (0.016) 0.042 (0.048 (0.035) 0.362 (0.171) 0.010 (0.017) (0.020) 0.025 (0.034 (0.025) 0.270 (0.292 (0.005 (0.124)) -0.013 (0.027) (0.015) (0.146) -0.020 (0.026) (0.124) (0.115) (0.015) (0.015) (0.146) (0.018) -0.020 (0.033) (0.029) (0.154) (0.154) (0.189) (0.016) (0.018) (0.016) (0.018) 0.007 (0.023) (0.017) (0.122) (0.132) (0.0132) (0.014) (0.017) 0.043 (0.051 (0.025) (0.158) (0.174) (0.014) (0.014) (0.015)	All (1) w/o Migrants (2) All (3) w/o Migrants (4) All (5) w/o Migrants (6) All (7) -0.034 (0.025) -0.045 (0.029) -0.025 (0.080 (0.137)) -0.014 (0.013) (0.016) -0.026 (0.024) 0.009 (0.029) (0.161) (0.137) (0.013) (0.016) (0.024) 0.009 (0.017) (0.020) (0.146) (0.145) (0.013) (0.016) (0.021) 0.042 (0.048 (0.035) 0.362 (0.144 (0.003) (0.017) 0.0010 (0.020) (0.031) 0.025 (0.034 (0.035) (0.155) (0.171) (0.017) (0.020) (0.031) 0.025 (0.026) (0.124) (0.115) (0.015) (0.146) (0.035) 0.027 (0.026) (0.124) (0.115) (0.015) (0.146) (0.035) 0.020 (0.033) (0.029) (0.154) (0.189) (0.016) (0.016) (0.018) (0.023) 0.007 (0.016 (0.023) (0.017) (0.122) (0.132) (0.132) (0.014) (0.014) (0.017) (0.023) 0.043 (0.031 (0.029) (0.158) (0.158) (0.174) (0.014) (0.014) (0.015) (0.024)	All (1) w/o Migrants (2) All (3) w/o Migrants (4) All (5) w/o Migrants (6) All (7) w/o Migrants (8) -0.034 -0.045 -0.025 0.080 -0.014 -0.019 -0.026 -0.031 (0.025) (0.029) (0.161) (0.137) (0.013) (0.016) (0.024) (0.026) 0.009 0.002 0.159 0.190 -0.002 -0.008 0.004 0.006 (0.017) (0.020) (0.146) (0.145) (0.013) (0.016) (0.021) (0.026) 0.042 0.048 0.362 0.414 0.003 -0.010 0.034 0.048 (0.039) (0.035) (0.155) (0.171) (0.017) (0.020) (0.031) (0.037) 0.025 0.034 0.270 0.292 -0.005 -0.013 -0.001 0.002 (0.027) (0.026) (0.124) (0.115) (0.015) (0.146) (0.035) (0.042) -0.020 -0.009 0.302	Positive SE Income Log SE Income Employed Log Hours (All Jobs) Salar All w/o Migrants (1) All w/o Migrants (1) 0.0031 0.0031 0.0066 0.0051

Notes: The table reports growing region/year interactions. Estimates for monthly wages omit data for 2000. Standard errors adjusted for department clustering are reported in parentheses.

TABLE 4b. ADULT LABOR MARKET OUTCOMES WITH POOLED INTERACTION TERMS

		Male and Fe	male Workers				Men	Only		
	Positive S	E Income	Log SE	Income	Emp	loyed	Log Hours	(All Jobs)	•	y Wage and Earnings
Interaction	No Trends	w/Trends	No Trends	w/Trends	No Trends	w/Trends	No Trends	w/Trends	No Trends	w/Trends
Terms	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
				Panel A: Gr	owing Effects (1	non-DMZ)				
1995-1997	0.038	0.005	0.251	0.288	0.004	0.008	0.020	0.013	-0.005	-0.035
	(0.020)	(0.020)	(0.121)	(0.162)	(0.010)	(0.014)	(0.018)	(0.031)	(0.037)	(0.058)
1998-2000	0.030	-0.035	0.260	0.333	0.013	0.020	0.044	0.030	0.079	0.026
	(0.021)	(0.035)	(0.130)	(0.286)	(0.012)	(0.025)	(0.017)	(0.057)	(0.044)	(0.122)
Trends		0.011		-0.012		-0.001		0.002		0.010
		(0.008)		(0.048)		(0.003)		(0.010)		(0.022)
				Pane	el B: DMZ Effe	cts				
1995-1997	0.002	-0.027	0.625	0.383	-0.020	0.058	0.050	0.057	-0.047	-0.067
	(0.025)	(0.045)	(0.125)	(0.150)	(0.040)	(0.044)	(0.032)	(0.084)	(0.017)	(0.105)
1998-2000	-0.097	-0.152	0.349	-0.129	-0.039	0.109	0.126	0.138	0.058	0.024
	(0.069)	(0.108)	(0.117)	(0.306)	(0.027)	(0.035)	(0.025)	(0.087)	(0.067)	(0.225)
Trends		0.009		0.077		-0.025		-0.002		0.006
		(0.007)		(0.057)		(0.004)		(0.018)		(0.031)
N	148,	306	40,	338	74,	781	69,	144	34,	451

Notes: The table reports pooled growing region/year interactions terms estimated using equation (2) in the text. Estimates for monthly wages omit data for 2000. Standard errors adjusted for department clustering are reported in parentheses.

TABLE 5a. OUTCOMES FOR CHILDREN - YEARLY INTERACTIONS, POOLED GROWING & DMZ

		Enrol	lment		I	abor Market	(Teenage Boy	/s)
	В	oys	G	irls	Emplo	oyment	Log Hour	s (All Jobs)
		w/o		w/o		w/o		w/o
Interaction	All	Migrants	All	Migrants	All	Migrants	All	Migrants
Terms	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
1994	-0.004	-0.042	0.054	0.044	0.021	0.025	-0.028	-0.040
	(0.042)	(0.050)	(0.027)	(0.038)	(0.048)	(0.055)	(0.082)	(0.107)
1995	-0.011	-0.012	0.061	0.049	-0.037	-0.048	0.069	0.074
1,,,,,	(0.031)	(0.037)	(0.032)	(0.031)	(0.038)	(0.039)	(0.085)	(0.099)
	(0.001)	(0.027)	(0.002)	(0.051)	(0.020)	(0.02)	(0.000)	(0.055)
1996	-0.006	0.001	0.025	0.023	0.033	0.034	0.187	0.173
	(0.035)	(0.033)	(0.031)	(0.035)	(0.052)	(0.058)	(0.082)	(0.071)
1997	-0.065	-0.073	0.031	0.025	0.078	0.092	0.081	0.041
	(0.037)	(0.035)	(0.043)	(0.047)	(0.040)	(0.046)	(0.069)	(0.076)
1998	-0.020	-0.042	0.036	0.042	0.010	0.001	0.228	0.243
1770	(0.046)	(0.051)	(0.062)	(0.074)	(0.061)	(0.069)	(0.062)	(0.062)
	(0.010)	(0.051)	(0.002)	(0.071)	(0.001)	(0.00))	(0.002)	(0.002)
1999	-0.055	-0.052	0.017	0.009	0.099	0.123	0.180	0.160
	(0.035)	(0.039)	(0.046)	(0.054)	(0.052)	(0.055)	(0.047)	(0.051)
2000	-0.054	-0.059	0.034	0.017	0.078	0.072	0.272	0.253
	(0.047)	(0.053)	(0.044)	(0.048)	(0.066)	(0.067)	(0.045)	(0.057)
N	27 202	22 605	25 771	21.250	22.265	10.210	10.500	10.104
N	27,382	22,695	25,771	21,259	22,365	18,319	12,528	10,104

Notes: The table reports growing region/year interactions. Standard errors adjusted for department clustering are reported in parentheses.

TABLE 5b. OUTCOMES FOR CHILDREN WITH POOLED INTERACTION TERMS

		Enro	llment		L	abor Market	(Teenage Boys	s)
	Во	oys		rls	Emplo	oyment	Log Hours	(All Jobs)
Interaction	No Trends	w/Trends	No Trends	w/Trends	No Trends	w/Trends	No Trends	w/Trends
Terms	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
			Panel A: Gro	wing Effects	(non-DMZ)			
1995-1997	-0.010	0.040	0.035	0.024	0.016	-0.087	0.112	0.117
	(0.023)	(0.038)	(0.016)	(0.042)	(0.032)	(0.046)	(0.039)	(0.058)
1998-2000	-0.028	0.069	0.027	0.006	0.073	-0.132	0.236	0.246
	(0.039)	(0.067)	(0.042)	(0.083)	(0.057)	(0.066)	(0.046)	(0.136)
Trends		-0.016		0.004		0.034		-0.002
		(0.011)		(0.015)		(0.010)		(0.020)
N	27,382	27,382	25,771	25,771	22,365	22,365	12,528	12,528
			Panel	B: DMZ Eff	ects			
1995-1997	-0.103 (0.024)	0.174 (0.030)	-0.115 (0.069)	-0.197 (0.076)	-0.032 (0.030)	-0.248 (0.171)	0.238 (0.039)	-0.004 (0.116)
1998-2000	-0.079	0.436	-0.138	-0.293	-0.213	-0.613	0.232	-0.216
1770-2000	(0.031)	(0.046)	(0.056)	(0.099)	(0.069)	(0.394)	(0.042)	(0.239)
	(0.051)	(0.010)	(0.020)	(0.0))	(0.00))	(0.5) 1)	(0.012)	(0.237)
Trends		-0.087		0.025		0.066		0.071
		(0.008)		(0.013)		(0.054)		(0.040)
N	27,	382	25,	771	22,	365	12,5	528

Notes: The table reports pooled growing region/year interactions terms estimated using equation (2) in the text. Estimates for monthly wages omit data for 2000. Standard errors adjusted for department clustering are reported in parentheses.

TABLE 6. ADULT AND TEENAGE LABOR MARKET OUTCOMES - RURAL AND URBAN EFFECTS

		Male and Fer	nale Workers			Men	Only		Teena	ge Boys
	Positive S	SE Income	Log SE	Income	_	Hours Jobs)	-	ly Wage and Earnings	_	Hours Jobs)
Interaction	14 growing	no med. prod.	14 growing	no med. prod.	14 growing	no med. prod.	14 growing	no med. prod.	14 growing	no med. prod.
Terms	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
				Par	nel A: Rural Eff	ects				
1995-1997	0.005	0.015	0.286	0.322	0.037	0.058	-0.019	-0.042	0.119	0.117
	(0.016)	(0.023)	(0.129)	(0.138)	(0.026)	(0.028)	(0.036)	(0.042)	(0.052)	(0.059)
1998-2000	-0.025	-0.022	0.295	0.298	0.079	0.122	0.045	-0.011	0.216	0.198
	(0.021)	(0.031)	(0.183)	(0.213)	(0.042)	(0.036)	(0.071)	(0.070)	(0.105)	(0.128)
				Par	nel B: Urban Eff	ects				
1995-1997	0.011 (0.018)	-0.01 (0.022)	0.155 (0.113)	0.198 (0.104)	0.025 (0.021)	0.041 (0.019)	-0.038 (0.043)	-0.060 (0.044)	-0.004 (0.068)	-0.014 (0.076)
1998-2000	-0.015	-0.052	0.174	0.197	0.040	0.074	-0.087	-0.122	-0.047	-0.017
	(0.030)	(0.035)	(0.184)	(0.189)	(0.041)	(0.040)	(0.071)	(0.077)	(0.091)	(0.109)
Trend	0.008	0.011	-0.004	-0.019	-0.005	-0.013	0.011	0.021	0.001	-0.004
	(0.005)	(0.006)	(0.028)	(0.032)	(0.007)	(0.006)	(0.012)	(0.014)	(0.015)	(0.019)
N	482,053	457,855	116,896	109,693	192,840	181,882	101,779	96,797	22,141	20,544

Notes: The table reports results from a stacked urban and rural sample. Estimates for monthly wages omit data for 2000. Standard errors adjusted for department clustering are reported in parentheses. Columns labelled "14 growing" use the same growing region as in Tables 4 and 5. Columns labelled "no medium producers," drop the 5 medium producing departments from the analysis.

TABLE 7. MORTALITY ESTIMATES

		14 Growing	Departments		No Medium Producers					
	No tr		With t	trends	No ti			trends		
	<i>C</i> :	DIVE	G :	DIVE	G :	DIVE	G :	DMZ		
	Growing	DMZ	Growing (3)	DMZ	Growing	DMZ (6)	Growing	DMZ		
	(1)	(2)	(3)	(4)	(5)	(0)	(7)	(8)		
				A. Unweighte	ed					
1993	0.026	-0.059			0.012	-0.059				
1993	(0.131)	(0.079)			(0.217)	(0.080)				
1994	0.123	-0.078	0.133	-0.145	0.209	-0.077	0.255	-0.145		
1771	(0.144)	(0.127)	(0.144)	(0.087)	(0.226)	(0.128)	(0.168)	(0.087)		
1995	0.102	0.198	0.118	0.098	0.236	0.198	0.302	0.097		
1,,,,	(0.131)	(0.279)	(0.203)	(0.175)	(0.190)	(0.280)	(0.284)	(0.176)		
1996	0.154	0.266	0.176	0.134	0.289	0.266	0.374	0.132		
1,,,0	(0.163)	(0.222)	(0.298)	(0.154)	(0.202)	(0.223)	(0.459)	(0.155)		
1997	0.230	0.402	0.259	0.237	0.346	0.403	0.451	0.236		
1,,,,	(0.178)	(0.293)	(0.344)	(0.162)	(0.225)	(0.294)	(0.501)	(0.163)		
1998	0.200	0.514	0.235	0.315	0.506	0.514	0.631	0.314		
1770	(0.200)	(0.288)	(0.452)	(0.171)	(0.243)	(0.290)	(0.727)	(0.173)		
1999	0.197	0.340	0.238	0.109	0.259	0.340	0.403	0.107		
1777	(0.119)	(0.243)	(0.482)	(0.203)	(0.155)	(0.244)	(0.763)	(0.204)		
2000	0.289	0.142	0.337	-0.122	0.358	0.142	0.522	-0.124		
2000	(0.120)	(0.251)	(0.550)	(0.246)	(0.144)	(0.252)	(0.874)	(0.248)		
Dept. trend	(0.120)	(0.231)	-0.006	0.033	(0.144)	(0.232)	-0.020	0.033		
Dept. trent			(0.064)	(0.040)			(0.103)	(0.041)		
			(0.001)	(0.010)			(0.103)	(0.011)		
				B. Weighted	I					
1993	-0.080	-0.100			-0.202	-0.101				
1775	(0.085)	(0.055)			(0.071)	(0.055)				
1994	0.045	-0.137	0.097	-0.105	0.062	-0.137	0.283	-0.105		
1///	(0.111)	(0.131)	(0.131)	(0.097)	(0.134)	(0.132)	(0.106)	(0.097)		
1995	0.141	0.126	0.205	0.159	0.183	0.125	0.473	0.159		
1775	(0.120)	(0.242)	(0.157)	(0.211)	(0.161)	(0.243)	(0.120)	(0.212)		
1996	0.163	0.178	0.241	0.214	0.180	0.178	0.538	0.214		
1770	(0.204)	(0.230)	(0.221)	(0.199)	(0.241)	(0.232)	(0.205)	(0.200)		
1997	0.179	0.260	0.269	0.298	0.220	0.259	0.647	0.298		
1771	(0.184)	(0.249)	(0.226)	(0.205)	(0.217)	(0.250)	(0.194)	(0.207)		
1998	0.120	0.434	0.222	0.474	0.169	0.434	0.665	0.475		
1996	(0.151)	(0.255)	(0.261)	(0.235)	(0.156)	(0.256)	(0.246)	(0.236)		
1999	0.313	0.233)	0.428	0.339	0.136)	0.296	0.240)	0.339		
1777	(0.149)	(0.205)	(0.324)	(0.223)	(0.185)	(0.206)	(0.268)	(0.224)		
2000	0.388	0.205)	0.516	0.223)	0.185)	0.206)	1.034	0.224)		
200U		(0.119				(0.118)	(0.288)			
Don't 4 1	(0.127)	(0.197)	(0.344)	(0.246)	(0.150)	(0.198)		(0.247)		
Dept. trend			-0.013	-0.002			-0.069 (0.034)	-0.002		
			(0.039)	(0.024)			(0.034)	(0.024)		

TABLE 8. MORTALITY ESTIMATES BY URBAN/RURAL RESIDENCE

		14 Growing	Departments		Without Medium Producers					
	No tr	rends	With	rends	No ti	rends	With	trends		
	Growing (1)	DMZ (2)	Growing (3)	DMZ (4)	Growing (5)	DMZ (6)	Growing (7)	DMZ (8)		
				A. Rural						
1994	0.038 (0.173)	-0.087 (0.094)	0.644 (0.418)	0.322 (0.129)	0.155 (0.186)	-0.086 (0.094)	0.645 (0.379)	0.321 (0.130)		
1995	0.310 (0.148)	0.378 (0.152)	1.305 (0.567)	1.057 (0.169)	0.301 (0.189)	0.378 (0.153)	1.111 (0.583)	1.055 (0.170)		
1996	0.356 (0.185)	0.580 (0.080)	1.740 (0.801)	1.529 (0.364)	0.317 (0.190)	0.580 (0.081)	1.446 (0.866)	1.527 (0.366)		
1997	0.397 (0.221)	0.283 (0.292)	2.170 (1.06)	1.503 (0.316)	0.301 (0.223)	0.284 (0.294)	1.751 (1.26)	1.501 (0.317)		
1998	0.266 (0.244)	0.491 (0.344)	2.427 (1.25)	1.981 (0.364)	0.324 (0.222)	0.491 (0.346)	2.094 (1.45)	1.978 (0.366)		
1999	0.415 (0.245)	0.055 (0.290)	2.965 (1.52)	1.816 (0.452)	0.329 (0.261)	0.056 (0.291)	2.420 (1.78)	1.813 (0.454)		
2000	0.506 (0.280)	-0.019 (0.223)	3.444 (1.76)	2.011 (0.606)	0.299 (0.263)	-0.019 (0.224)	2.709 (2.00)	2.008 (0.610)		
Dept. trend			-0.388 (0.217)	-0.270 (0.094)			-0.320 (0.245)	-0.270 (0.094)		
				B. Urban						
1994	0.196 (0.129)	-0.048 (0.092)	0.147 (0.295)	0.183 (0.226)	0.279 (0.134)	-0.048 (0.093)	0.246 (0.384)	0.183 (0.227)		
1995	0.052 (0.130)	0.100 (0.237)	-0.030 (0.428)	0.484 (0.371)	0.200 (0.172)	0.099 (0.274)	0.146 (0.586)	0.483 (0.374)		
1996	0.165 (0.155)	-0.040 (0.333)	0.051 (0.531)	0.498 (0.474)	0.345 (0.196)	-0.041 (0.335)	0.269 (0.727)	0.496 (0.477)		
1997	0.109 (0.181)	0.146 (0.292)	-0.038 (0.687)	-0.838 (0.573)	0.356 (0.188)	0.146 (0.294)	0.258 (0.887)	0.837 (0.576)		
1998	0.034 (0.192)	0.332 (0.266)	-0.145 (0.825)	1.177 (0.655)	0.333 (0.230)	0.331 (0.268)	0.213 (1.09)	1.176 (0.659)		
1999	0.053 (0.165)	0.329 (0.259)	-0.158 (0.999)	1.328 (0.744)	0.217 (0.203)	0.328 (0.260)	0.076 (1.37)	1.326 (0.749)		
2000 Dont trand	0.080 (0.141)	0.115 (0.306)	-0.164 (1.16)	1.268 (0.888)	0.246 (0.170)	0.115 (0.306)	0.083 (1.59)	1.266 (0.894)		
Dept. trend			0.033 (0.152)	-0.154 (0.119)			0.022 (0.204)	-0.154 (0.120)		

TABLE 9. MORTALITY ESTIMATES BY URBAN/RURAL RESIDENCE AND AGE

		14 Growing	Departments		No Medium Producers				
	No t	rends	With	trends	No t	rends	With	trends	
	Age<35 (1)	Age>=35 (2)	Age<35 (3)	Age>=35 (4)	Age<35 (5)	Age>=35 (6)	Age<35 (7)	Age>=35 (8)	
				A. Rural					
1994	0.072 (0.222)	-0.029 (0.157)	0.067 (0.263)	-0.034 (0.151)	0.299 (0.200)	-0.064 (0.169)	0.411 (0.228)	0.048 (0.176)	
1995-97	0.386 (0.156)	0.349 (0.146)	0.374 (0.245)	0.338 (0.180)	0.401 (0.181)	0.294 (0.129)	0.661 (0.294)	0.554 (0.197)	
1998-00	0.414 (0.202)	0.313 (0.244)	0.392 (0.309)	0.292 (0.249)	0.435 (0.206)	0.173 (0.229)	0.913 (0.323)	0.651 (0.262)	
Trend				003 050)				073 050)	
				B. Urban					
1994	0.165 (0.136)	0.140 (0.155)	0.146 (0.146)	0.121 (0.179)	0.282 (0.156)	0.112 (0.192)	0.289 (0.156)	0.119 (0.221)	
1995-97	0.071 (0.149)	0.122 (0.131)	0.028 (0.190)	0.079 (0.199)	0.263 (0.163)	0.215 (0.171)	0.278 (0.209)	0.230 (0.255)	
1998-00	0.123 (0.153)	0.070 (0.141)	0.043 (0.280)	-0.010 (0.282)	0.341 (0.154)	0.206 (0.169)	0.369 (0.348)	0.234 (0.368)	
Trend	` ,	` ,	0.0	012 (039)	` ,	` ,	-0.	004 049)	

TABLE 10. MORTALITY ESTIMATES BY PREVIOUS GUERILLA PRESENCE

		14 Growing	Departments			No Mediur	n Producers	
	No ti	rends	With	trends	No t	rends	With	trends
	low (1)	high (2)	low (3)	high (4)	low (5)	high (6)	low (7)	high (8)
	(1)	(=)				(0)	(,)	(6)
			Α	. By 1994 FA	RC			
1994	083	.116	088	.112	.269	.071	.380	.182
	(.288)	(.156)	(.317)	(.157)	(.361)	(.096)	(.391)	(.095)
1995-97	.073	.593	.064	.583	.277	.458	.537	.717
	(.191)	(.173)	(.281)	(.184)	(.138)	(.163)	(.256)	(.233)
1998-00	.017	.550	001	.532	.106	.351	.582	.829
	(.190)	(.338)	(.362)	(.279)	(.108)	(.331)	(.340)	(.310)
Trend			.0	03			0)73
			0.)	49)			0.)	50)
			В. Е	By 1995 FARC	/ELN			
1994	296	.208	303	.201	176	.268	068	.375
	(.158)	(.176)	(.175)	(.202)	(.116)	(.190)	(.128)	(.210)
1995-97	.271	.477	.254	.460	.256	.441	.508	.689
	(.093)	(.189)	(.211)	(.230)	(.098)	(.185)	(.218)	(.266)
1998-00	.073	.633	.043	.603	.087	.514	.547	.974
	(.272)	(.242)	(.394)	(.283)	(.304)	(.219)	(.433)	(.282)
Trend				05)71
			0.)	49)			0.)	49)

TABLE A1. ROBUSTNESS CHECKS FOR LABOR MARKET OUTCOMES

	Male and Female Workers				Men Only				Teenage Boys	
	Positive SE Income		Log SE Income		Log Hours		Log Monthly Wage and		Log Hours	
		no med.		no med.		no med.	•	no med.	'	no med.
Interaction	14 growing	prod.	14 growing	prod.	14 growing	prod.	14 growing	prod.	14 growing	prod.
Terms	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
				Panel A: R	ainfall Control, N	lo Trends				
1995-1997	0.037	0.060	0.234	0.249	0.019	0.017	-0.001	0.017	0.107	0.112
	(0.021)	(0.026)	(0.119)	(0.162)	(0.017)	(0.024)	(0.036)	(0.055)	(0.039)	(0.037)
1998-2000	0.033	0.058	0.256	0.177	0.039	0.034	0.082	0.086	0.232	0.179
	(0.021)	(0.021)	(0.126)	(0.115)	(0.017)	(0.018)	(0.041)	(0.057)	(0.045)	(0.033)
			Panel B: Ra	ainfall Conti	ol and Linear De	partment T	ype Trend			
1995-1997	0.004	-0.004	0.284	0.287	0.009	0.015	-0.026	-0.054	0.123	0.121
	(0.018)	(0.014)	(0.158)	(0.155)	(0.030)	(0.025)	(0.059)	(0.062)	(0.059)	(0.079)
1998-2000	-0.032	0.068	0.357	0.253	0.018	0.03	0.037	-0.043	0.265	0.197
	(0.035)	(0.022)	(0.289)	(0.307)	(0.056)	(0.032)	(0.131)	(0.149)	(0.140)	(0.176)
Trend	0.011	0.021	-0.017	-0.013	0.003	0.001	0.008	0.024	-0.006	-0.003
	(0.007)	(0.006)	(0.049)	(0.048)	(0.010)	(0.006)	(0.024)	(0.031)	(0.021)	(0.027)
				Panel	C: Initial GSP T	rend				
1995-1997	0.031	0.054	0.246	0.260	0.029	0.028	-0.017	0.001	0.108	0.110
	(0.023)	(0.030)	(0.137)	(0.177)	(0.023)	(0.031)	(0.039)	(0.053)	(0.044)	(0.045)
1998-2000	0.022	0.049	0.279	0.193	0.057	0.055	0.050	0.051	0.234	0.169
	(0.025)	(0.029)	(0.171)	(0.179)	(0.013)	(0.036)	(0.059)	(0.076)	(0.062)	(0.060)
Trend	-0.003	-0.002	0.007	0.004	0.005	0.005	-0.007	-0.007	0.001	-0.002
	(0.004)	(0.004)	(0.028)	(0.029)	(0.005)	(0.005)	(0.010)	(0.010)	(0.009)	(0.009)
				Panel D: In	itial FARC Prese	nce Trend				
1995-1997	0.034	0.059	0.283	0.268	0.016	0.015	-0.003	0.018	0.133	0.117
	(0.022)	(0.026)	(0.125)	(0.154)	(0.020)	(0.026)	(0.037)	(0.053)	(0.045)	(0.040)
1998-2000	0.034	0.059	0.256	0.171	0.040	0.035	0.075	0.079	0.233	0.177
	(0.021)	(0.021)	(0.130)	(0.117)	(0.018)	(0.019)	(0.043)	(0.062)	(0.047)	(0.033)
Trend	-0.002	-0.001	0.023	0.026	-0.002	-0.003	0.000	0.001	0.010	0.004
	(0.001)	(0.001)	(0.007)	(0.005)	(0.002)	(0.002)	(0.001)	(0.001)	(0.004)	(0.003)
N	148,306	126,358	40,338	34,111	69,144	58,314	34,451	29,275	12,528	10,685

Notes: The table reports estimates from models and samples similar to those used to construct the estimates in Tables 4b and 5b, with the addition of trend and control variables as indicated.

TABLE A2. ADDITIONAL MORTALITY ESTIMATES

		14 Growing	Departments		No Medium Producers					
	No trends		With trends		No trends		With trends			
	No ctls (1)	Rainfall (2)	No ctls (3)	Rainfall (4)	No ctls (5)	Rainfall (6)	No ctls (7)	Rainfall (8)		
	(1)	(2)	(3)	(4)	(3)	(0)	(7)	(0)		
				A. Unweighte	d					
1993	025	025	024	023	017	022	014	018		
	(.158)	(.146)	(.157)	(.145)	(.162)	(.147)	(.161)	(.146)		
1994	.129	.111	.131	.113	.136	.115	.141	.120		
	(.174)	(.148)	(.175)	(.149)	(.180)	(.152)	(.182)	(.155)		
1995	.248	.227	.251	.230	.237	.216	.243	.222		
	(.157)	(.145)	(.158)	(.146)	(.164)	(.152)	(.164)	(.153)		
1996	.283	.275	.286	.278	.281	.268	.289	.276		
	(.163)	(.158)	(.165)	(.160)	(.175)	(.169)	(.178)	(.172)		
1997	.305	.316	.309	.320	.315	.325	.325	.335		
	(.182)	(.194)	(.182)	(.194)	(.192)	(.202)	(.191)	(.202)		
1998	.536	.535	.540	.540	.481	.479	.492	.491		
	(.201)	(.204)	(.200)	(.203)	(.203)	(.206)	(.202)	(.204)		
1999	.251	.254	.256	.260	.281	.282	.293	.296		
	(.135)	(.135)	(.137)	(.138)	(.142)	(.141)	(.146)	(.145)		
2000	.240	.234	.245	.241	.289	.277	.303	.291		
	(.134)	(.138)	(.139)	(.144)	(.141)	(.144)	(.150)	(.154)		
Trend			.002	.002			.003	.003		
			(.005)	(.005)			(.005)	(.005)		
				B. Weighted						
1993	199	198	205	203	184	184	188	187		
	(.060)	(.064)	(.062)	(.066)	(.068)	(.071)	(.069)	(.072)		
1994	.022	.021	.014	.013	.026	.026	.021	.020		
	(.115)	(.113)	(.117)	(.116)	(.127)	(.125)	(.130)	(.128)		
1995	.157	.156	.146	.145	.173	.173	.166	.166		
	(.138)	(.138)	(.138)	(.138)	(.147)	(.147)	(.148)	(.147)		
1996	.150	.151	.136	.137	.178	.178	.169	.169		
	(.199)	(.200)	(.200)	(.201)	(.222)	(.222)	(.223)	(.224)		
1997	.199	.199	.182	.183	.222	.223	.211	.212		
	(.186)	(.187)	(.191)	(.192)	(.195)	(.196)	(.199)	(.200)		
1998	.203	.201	.183	.182	.213	.212	.200	.199		
	(.146)	(.146)	(.154)	(.153)	(.150)	(.149)	(.154)	(.154)		
1999	.309	.311	.287	.289	.356	.357	.341	.342		
	(.159)	(.160)	(.164)	(.165)	(.167)	(.168)	(.172)	(.172)		
2000	.276	.276	.251	.251	.351	.350	.334	.334		
	(.147)	(.146)	(.153)	(.152)	(.144)	(.144)	(.153)	(.153)		
Trend	(17)	()	009	009	()	(1 1)	005	005		
			(.010)	(.010)			(.009)	(.009)		

Notes: The table reports estimates from models and samples similar to used to construct the estimates in Table 7, with the addition of trend and control variables as indicated.

