

Income Taxation and State Capacities in Chile: Measuring Institutional Development Using Historical Earthquake Data

Hector Bahamonde • Postdoctoral Fellow • Tulane University

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Overview

[Origins] Most theories emphasize how important fiscal development is for state consolidation. **However**, the origins of fiscal development are less clear.

[Measurement] Most theories provide **historical** explanations for state consolidation, **yet**, they lack of a **historical** measurement capable of capturing levels of state capacities overtime.

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- I find that these gaps represent important **theoretical** and **empirical deficits**.

Taxation and State Capacities

Convince you:

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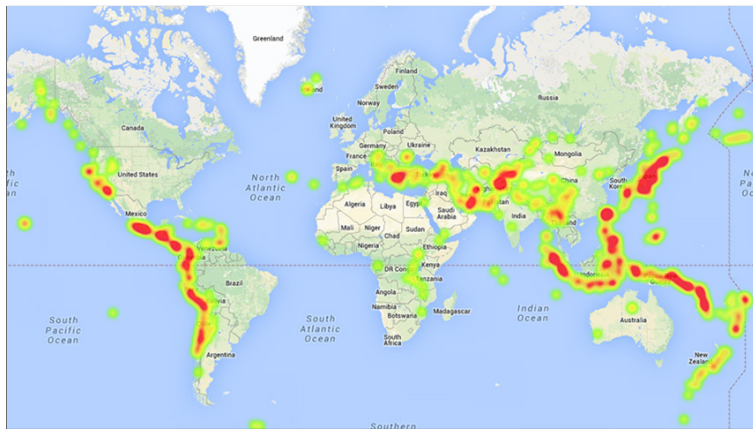
Taxation and State Capacities

Convince you:

1. Higher levels of sectoral competition promoted the implementation of the income tax.
2. The income tax fostered higher levels of state consolidation overtime.
3. Earthquake death-tolls are good proxies to measure state capacities.

Why Earthquakes?

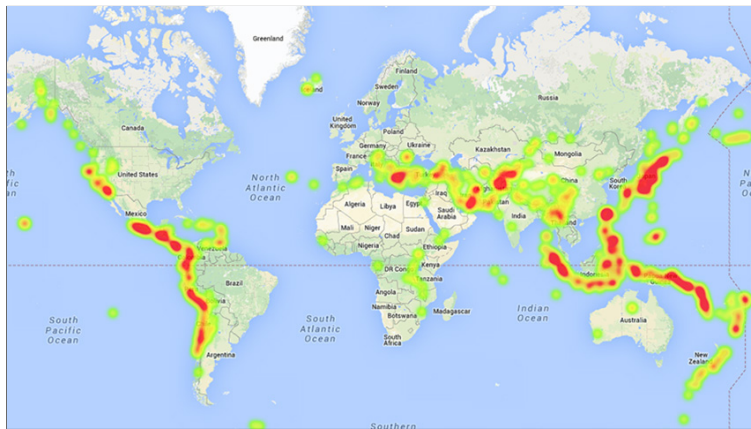
The **capacity** of the state to **enforce** quake-sensitive **building codes** throughout the territory, is a **reflection** of its **overall** state-capacities.



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Why Earthquakes?

Earthquakes are **exogenous** to regime type, levels of political/economic development, and other sources of variation.



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Earthquake and States Capacities

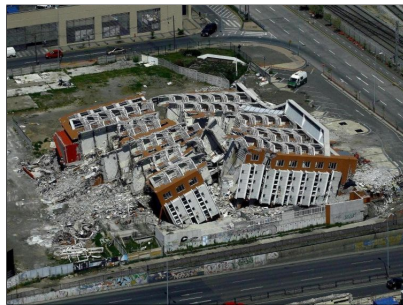
2010 Haiti: 7M, 100,000 casualties

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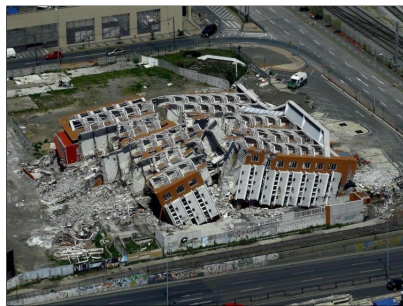
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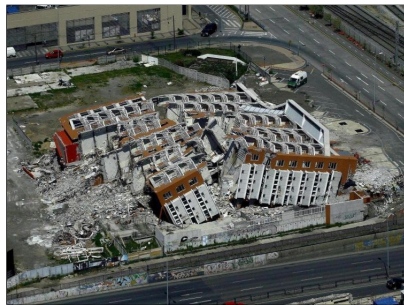
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Why did a weaker earthquake flatten Haiti? **Intuition:** Chile has better “state capacities” compared to Haiti.

The Importance of Building Codes

There exists a **popular** consensus on that **building codes** *do* reduce death tolls.



Dual Political Economy and Taxation

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 - [Agr] Since **land fixity** increases the risk premium of the landed elite’s main asset, landowners systematically **resist** taxation.
 - [Ind] As capital can be **reinvested in nontaxable sectors**, industrialists’ preferences toward taxation are more **elastic**.

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- Monitoring private incomes, and converting them into **public property**, *fostered* state development:
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- Income taxation generated positive **spillover effects** for state-making, rising **economies of scale** of the **operational efficiencies** of the bureaucracy.

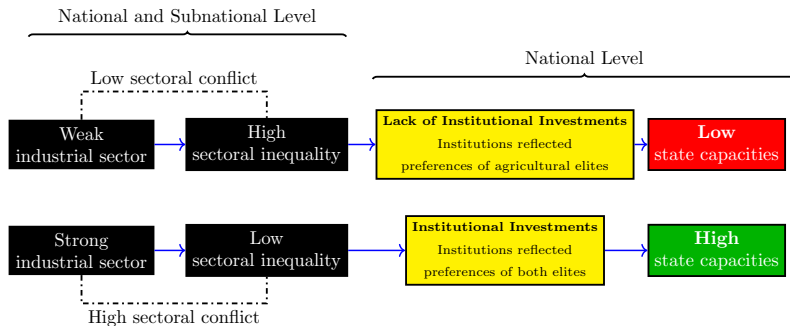
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In simple, the same bureaucracies that were sent to collect taxes, used the acquired knowledge to perform other state tasks.

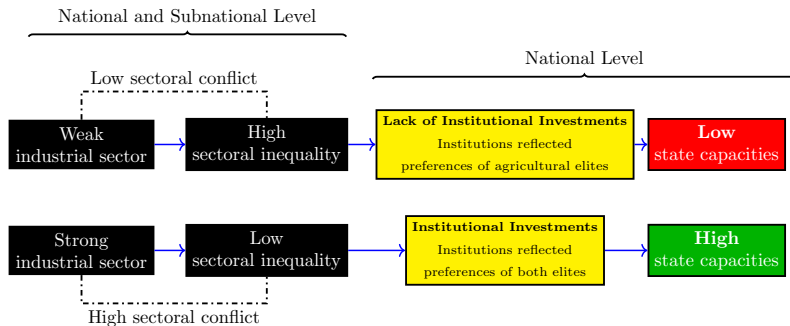
Argument

Higher levels of subnational and national **sectoral contestation** fostered state-capacities overtime.



Conceptualizing Contestation

Industrial expansion in the **regions**, challenged the **national/hegemonic agricultural political economy**.



Chilean Case Study

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- Industrial elites agreed to comply with the central level and pay the **income tax** in exchange of public goods delivered at the local level (bridges, ports, etc.).

The Theory Should Pass Two Tests

1. The state should have higher capacities (i.e. *lower death tolls*) when it's subnationally contested.
2. Implementation of the income tax should produce higher state-capacities (i.e. *lower death tolls*) overtime.

Analyses

Data Subnational and national data on Chile (1907 to 2012).

National Sectoral outputs to measure sectoral competition. **plot**

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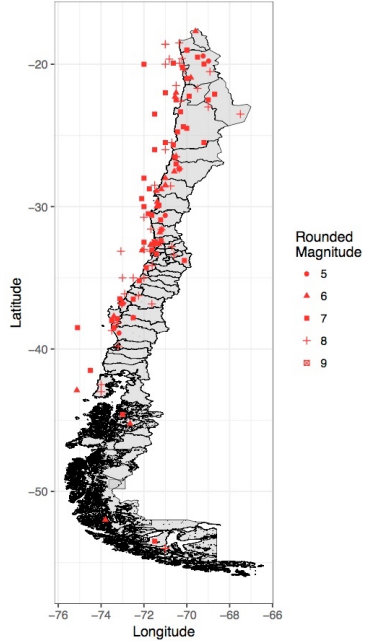
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Model Bayesian Hierarchical Poisson model with year fixed-effects, to account for the count of deaths. [Jags Code](#)



Subnational Contestation and State Development

Deaths $\sim \text{Poisson}(\lambda_i)$

Distribution of Deaths

$$\log(\lambda_i) = \mu + \beta_{1j} \text{Proportion}_{\text{Ind}}^{\text{Agr}} i + \beta_{2j} \text{Magnitude}_i^2 + \beta_3 \text{Latitude}_i + \beta_4 \text{Longitude}_i + \beta_5 \text{Population}_i + \beta_6 \text{Urban}_i + \beta_{7t} \text{Year}_i$$

where,

$i_{1,...,I}$ where $I = 91$ earthquakes

$j_{1,...,J}$ where $J = 3$ sectors @ subnational level (agr, ind, mixed)

$t_{1,...,T}$ where $T = 59$ years.

4 chains, 200K iterations, burn-in of 5000.

Densities

Trace plots

Model fit

Table

Download detailed diagnostics plots

Proportion

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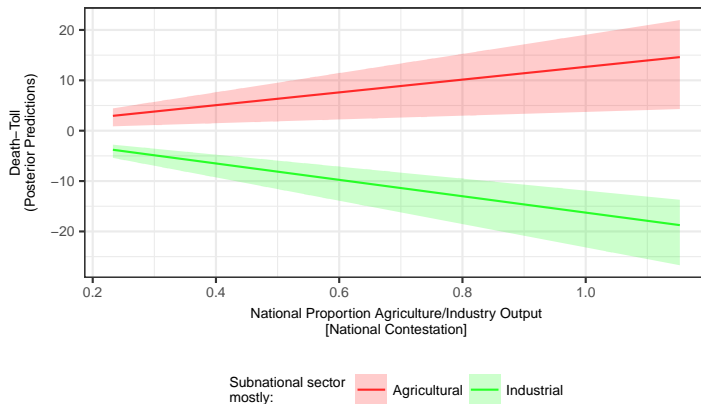
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Proportion

Subnational Contestation Increases State Capacities



Income Taxation and State Development Overtime

Deaths $\sim \text{Poisson}(\lambda_i)$

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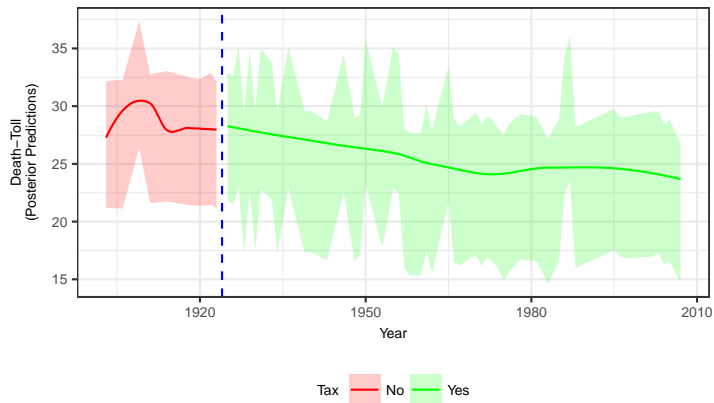
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Income Taxation Increase State Development Overtime



Summary

1. **Sectoral contestation**, between a strong national agriculture, and strong regional industrial clusters, helped to foster state consolidation.
2. **Income taxation** increased state capacities overtime.
3. **Limitations**: this rationale only applies for cases where there are earthquakes.

What's ahead:

- More data should be collected, and the mechanism should be examined in other countries of the region.

Thank you

www.HectorBahamonde.com

[TOC](#)

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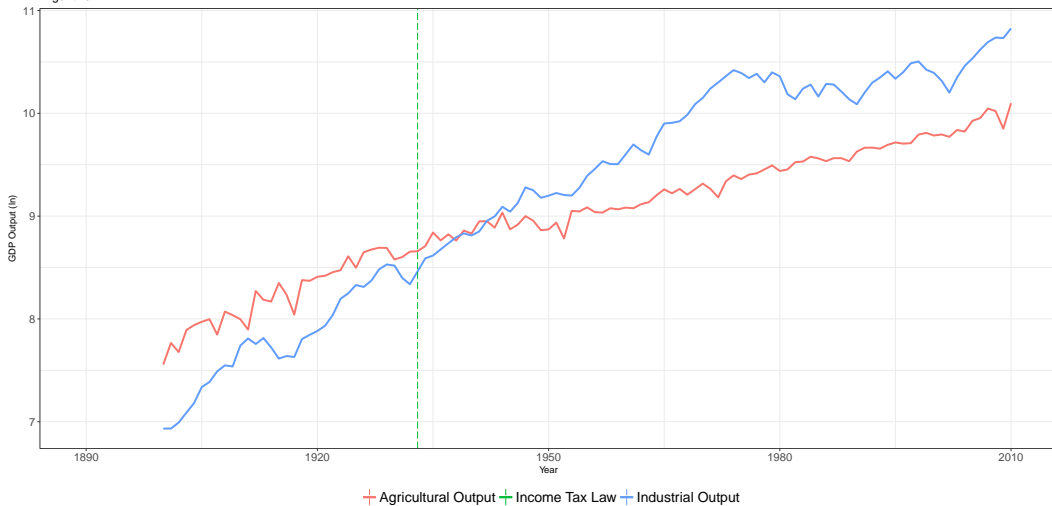
TOC

- P2: Unit Root Tests
- P2: Johansen Tests for Cointegration
- P2: Lags Tests
- P2: Sectoral Outputs
- P2: Granger-causality Tests
- P3: Sectoral Model Density Plots
- P3: Income Tax Model Density Plots
- P3: Sectoral Model Trace Plots
- P3: Income Tax Model Trace Plots
- P3: Sectoral and Income Tax Model Goodness of Fit Plot
- P3: Dependent Variable Agriculture
Industry
- P3: Sectoral Model Regression Table
- P3: Income Tax Model Regression
- P3: Jags code for sectoral model
- P3: Distribution of Deaths
- Credible Threats
- From Conflict to Cooperation
- War was in 1891, but income tax was implemented in 1924
- Why does taxation increase with sectoral competition?
- Everything depends on industrial expansion. Where does industry come from, then?
- Why not indirect taxation?
- Duration Models

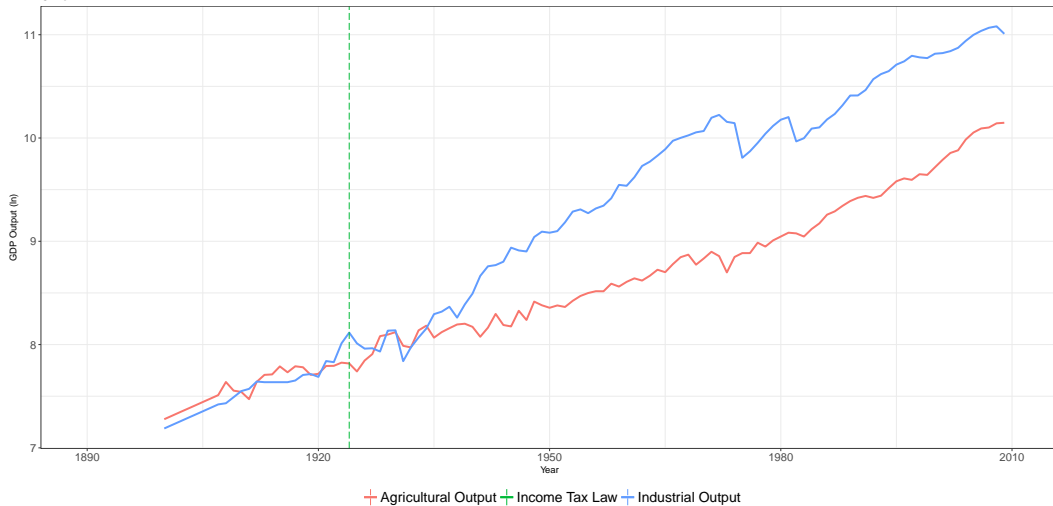
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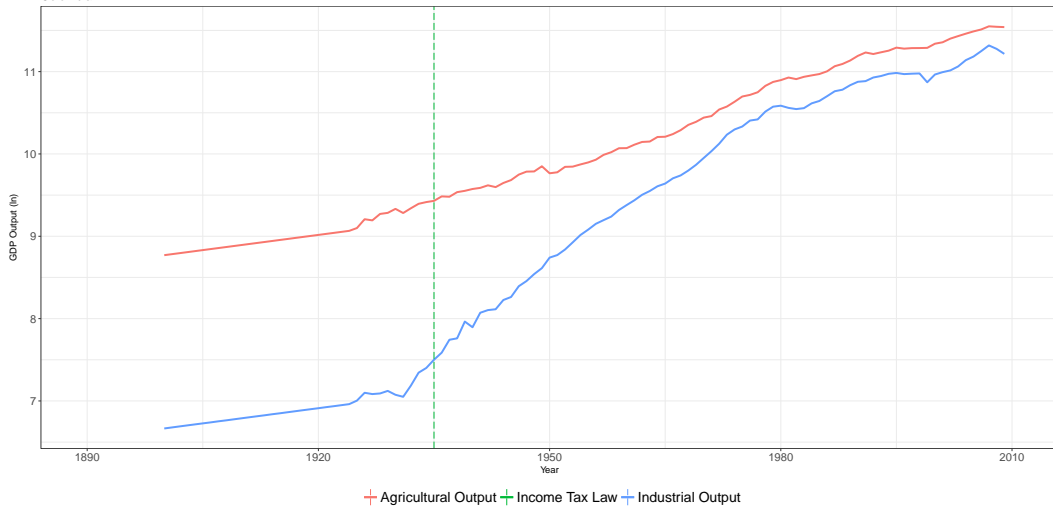
Argentina



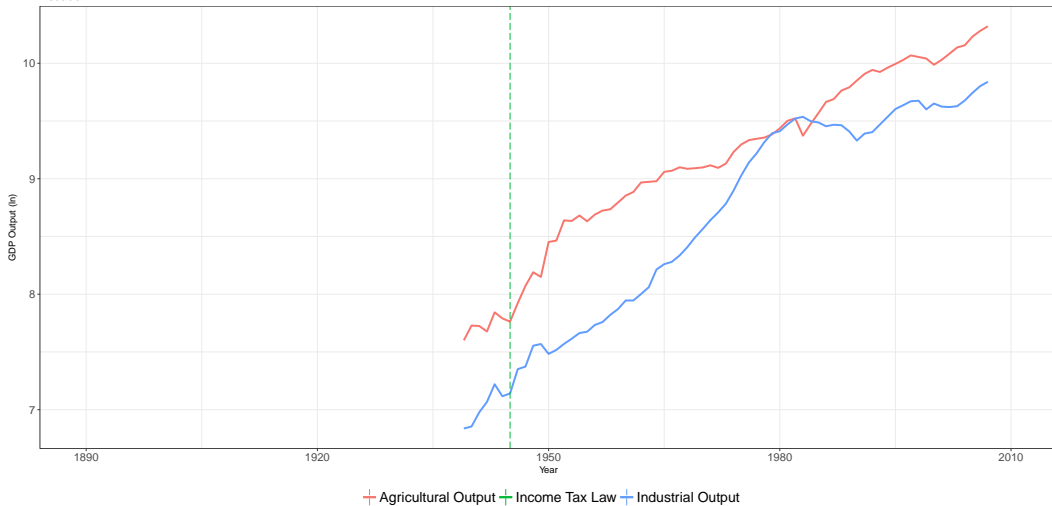
Chile



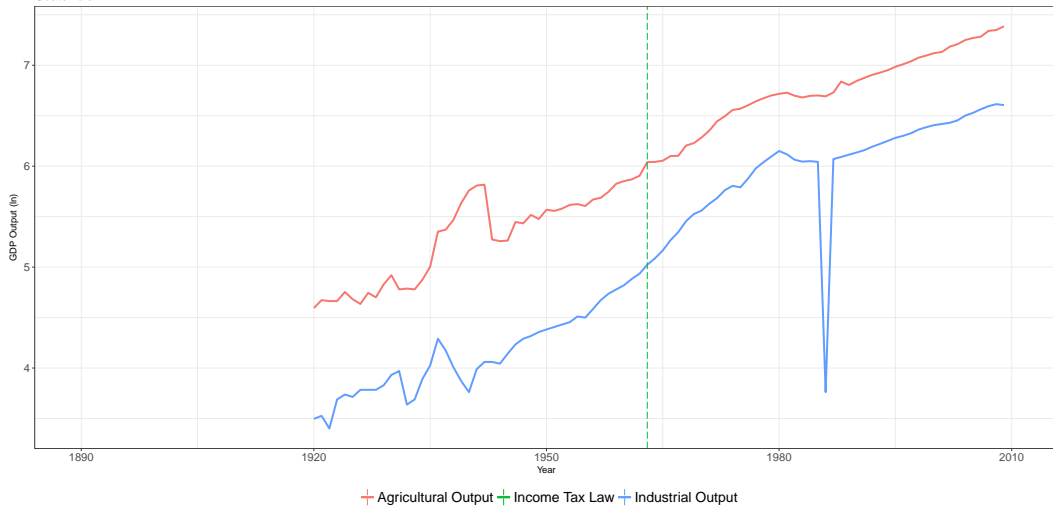
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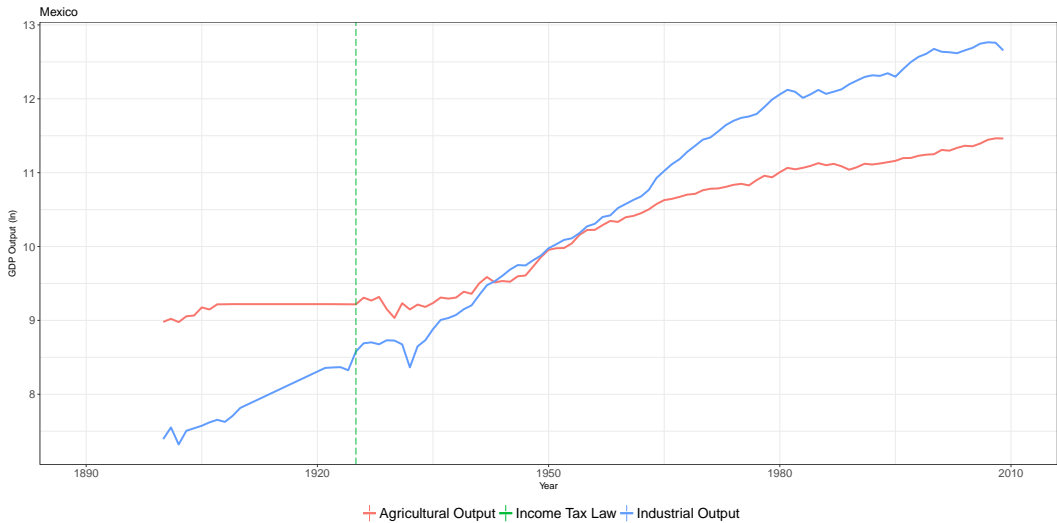


Ecuador

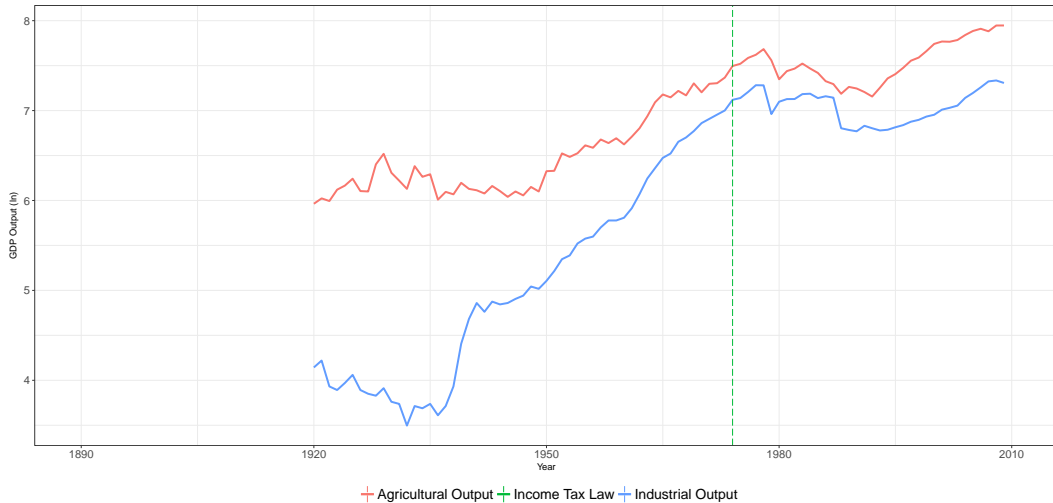


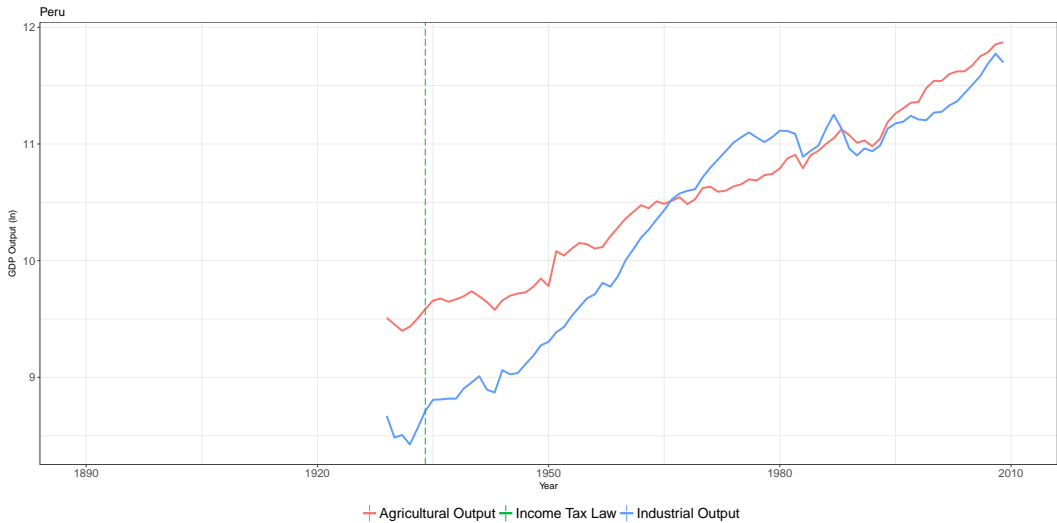
Guatemala



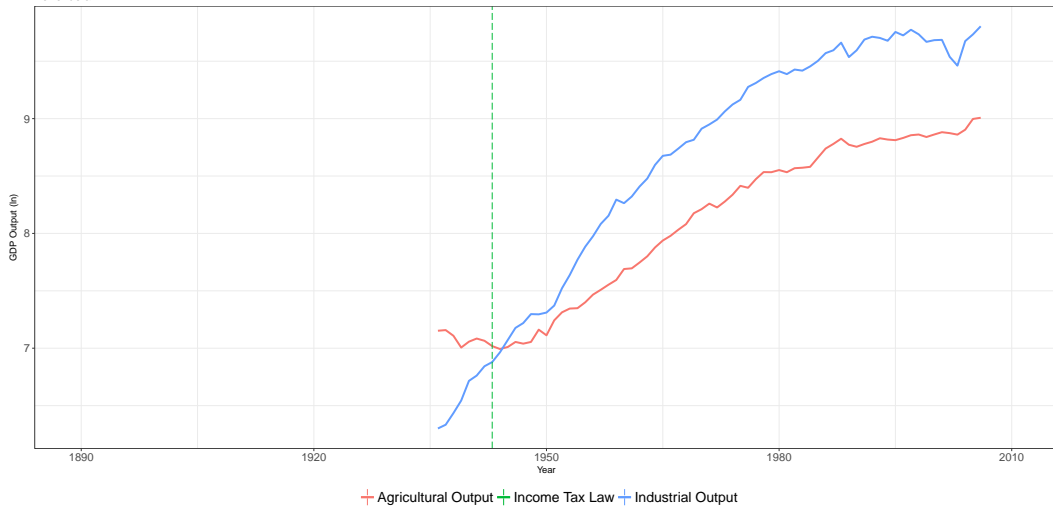


Nicaragua





Venezuela



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| Total Population | 0.000** (0.000) | | | | |
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| Total Population (ln) | | 5.059* (2.228) | 1.259 (1.052) | 1.030** (0.391) | 4.676 (2.682) |
| Manufacture Output (ln) | | | 1.924*** (0.514) | 0.668*** (0.143) | 7.148 (4.815) |
| Agricultural Output (ln) | | | -1.596** (0.603) | -0.941*** (0.281) | -6.465 (4.636) |
| AIC | 12.796 | 10.894 | | 4505.538 | 11.056 |
| R ² | 0.059 | 0.068 | | 0.341 | 0.065 |
| Max. R ² | 0.085 | 0.088 | | 0.997 | 0.085 |
| Num. events | 9 | 9 | | 610 | 9 |
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| Missings | 0 | 0 | | 0 | 0 |
| PH test | 0.388 | 0.877 | | | 0.667 |
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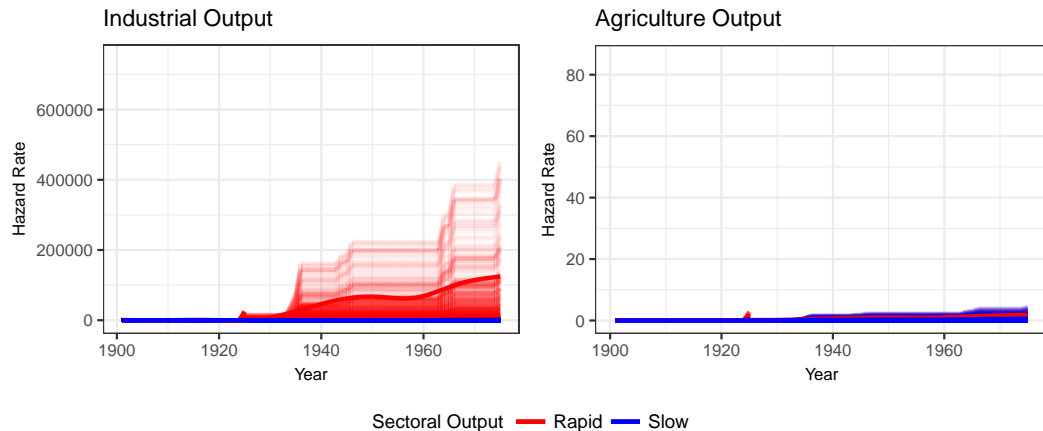
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| Manufacture Output (ln) | | | 1.924*** (0.514) | 0.668*** (0.143) | 7.148 (4.815) |
| Agricultural Output (ln) | | | -1.596** (0.603) | -0.941*** (0.281) | -6.465 (4.636) |
| AIC | 12.796 | 10.894 | | 4505.538 | 11.056 |
| R ² | 0.059 | 0.068 | | 0.341 | 0.065 |
| Max. R ² | 0.085 | 0.088 | | 0.997 | 0.085 |
| Num. events | 9 | 9 | | 610 | 9 |
| Num. obs. | 241 | 232 | 842 | 842 | 241 |
| Missings | 0 | 0 | | 0 | 0 |
| PH test | 0.388 | 0.877 | | | 0.667 |
| Num. clust. | | | 9 | | |

*** $p < 0.001$, ** $p < 0.01$, * $p < 0.05$, $p < 0.1$. Robust standard errors in all models

| | Cox (1 lag) | Cox (1 lag, ln) | Logit GEE | Conditional Logit (FE) | Spatial Dependence |
|---|--------------------|--------------------|---------------------|------------------------|-------------------------------|
| Manufacture Output _{t-1} | 4.923** (1.851) | | | | |
| Agricultural Output _{t-1} | -4.208* (1.638) | | | | |
| Total Population | 0.000** (0.000) | | | | |
| Manufacture Output _{t-1} (ln) | | 7.685* (3.333) | | | |
| Agricultural Output _{t-1} (ln) | | -6.971* (3.227) | | | |
| Total Population (ln) | | 5.059* (2.228) | 1.259 (1.052) | 1.030** (0.391) | 4.676 [·] (2.682) |
| Manufacture Output (ln) | | | 1.924*** (0.514) | 0.668*** (0.143) | 7.148 (4.815) |
| Agricultural Output (ln) | | | -1.596** (0.603) | -0.941*** (0.281) | -6.465 (4.636) |
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| Num. clust. | | | 9 | | |

*** $p < 0.001$, ** $p < 0.01$, * $p < 0.05$, [·] $p < 0.1$. Robust standard errors in all models

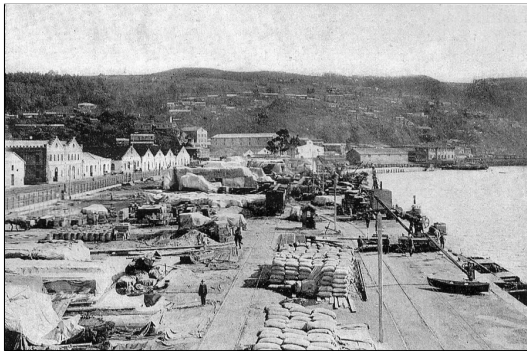
Simulated sectoral hazard rates of implementing the income tax law.
HR: probability that a case will fail at time t .



Why **Not** *Indirect* Taxation

Indirect taxes (like import taxes) require less **state efforts** to capture revenue.

Staffing an office, **waiting** for the ships to come in and **count** the goods. **Sacks of wheat**, for ex.



Talcahuano Port, Chile 19th Century.

```

1 | model.jags.sectoral <- function() {
2 |   for (i in 1:N){ # number of earthquakes
3 |
4 |     Deaths[i] ~ dpois(lambda[i]) log(lambda[i]) <~
5 |       b.propagmanu[Sector[i]]*propagmanu[i] + # multi-level
6 |       b.Magnitude[Sector[i]]*Magnitude[i] + # multi-level
7 |       b.p.Population*p.Population[i] +
8 |       b.Urban*Urban[i] +
9 |       b.year[yearID[i]] + # year fixed-effects
10 |       b.r.long*r.long[i] +
11 |       b.r.lat*r.lat[i] +
12 |       mu ## intercept
13 |   }
14 |
15 |   ## Non-Informative/Flat Priors
16 |   b.r.lat ~ dnorm(0, 0.01)
17 |   b.r.long ~ dnorm(0, 0.01)
18 |   mu ~ dnorm(0, 0.01) ## intercept
19 |   b.p.Population ~ dnorm(0, 0.01)
20 |   b.Urban ~ dnorm(0, 0.01)
21 |
22 |   ## Year Fixed-Effects
23 |   for (t in 1:yearN){
24 |     b.year[t] ~ dnorm(m.b.year[t], tau.b.year[t])
25 |     m.b.year[t] ~ dnorm(0, 0.01)
26 |     tau.b.year[t] ~ dgamma(0.5, 0.001) # uninformative Gamma priors
27 |   }
28 |
29 |   ## Varying Slopes for Magnitude (unmodeled)
30 |   for (k in 1:NSector){#
31 |     b.Magnitude[k] ~ dnorm(m.Magnitude[k], tau.Magnitude[k])
32 |     m.Magnitude[k] ~ dnorm(0, 0.01)
33 |     tau.Magnitude[k] ~ dgamma(0.5, 0.001) # uninformative Gamma priors
34 |   }
35 |
36 |   ## Varying Slopes for Agr/Ind Proportion (unmodeled)
37 |   for (k in 1:NSector){#
38 |     b.propagmanu[k] ~ dnorm(m.b.propagmanu[k], tau.b.propagmanu[k])
39 |     m.b.propagmanu[k] ~ dnorm(0, 0.01)
40 |     tau.b.propagmanu[k] ~ dgamma(0.5, 0.001) # uninformative Gamma priors
41 |   }
42 |
43 | }

```

Sectoral Competition and Taxation?

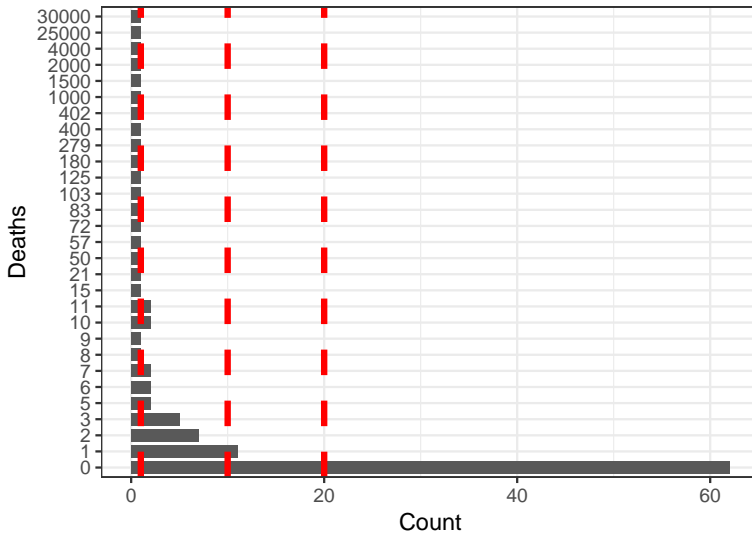
Agricultural production, as it needs mostly land, it does not rely on capital as much as the industrial sector does. Moreover, they oppose taxation because their main asset (land) is fixed, hence landowners not being able to move their asset, resist taxation. On the contrary, industrial elites rely on public goods that are beneficial for their business (railroads, bridges, etc.). And while industrialists would prefer imposing higher import taxes (NOT the income tax), that increases the price of importing industrial capital (for ex., machines). Consequently, their second best choice is imposing an income tax.

For these reasons, the emergence of the industrial sector (which implies higher levels of sectoral/elite contestation) leads to the implementation of the income tax.

Where does industry come from?

p. 20 of dissertation. Industry, as predicted by the dual sector model, came from agriculture:

- After the mining boom, mining elites shifted their focus to what is considered the first *true* industrial work which began under agricultural auspices: the cotton mills: “[t]he first power looms were brought [in Perú, Ecuador, and Venezuela] in the 1840s, 1850s; but in all three they were a failure, some of the early mills in Ecuador being destroyed by an earthquake. It was not until after 1890 that the textile industries of these nations began to operate with reasonable success. Guatemala’s first cotton mill was established in 1882, and between that date and 1910 a few mills appeared in Chile, Argentina, Uruguay, and Colombia.”
- The first industries were called *obrajes* and beyond textiles, early industrialists processed other agricultural goods. For example, animal grease and tallow, dried and cured meats, flour, bread, beer, wines and spirits, being most of them for domestic consumption. Sugar was used in the production of chocolate, candies and biscuits.
- The industrial sector was boosted by favorable international conditions, many times stimulating a positive complementarity between the two sectors. Industrial activities started very small, progressing “from the shop to the factory during the latter half of the nineteenth century.”
- Importantly, modern industrialization did *not* begin with ISI, but around 1900. Others find that the “fact that manufacturing was alive and thriving in Latin America before the 1929 crash is now beyond question.” And that the “development of large-scale, mechanized (and even “heavy”) industry can be dated back to the 1890s.” By the 1870’s the carriage industry was on a firm basis.



But Where Does Industrialization Come From?!

The theory puts heavy emphasis on the role of industrialization on state development. However, *Where does industrialization come from?*

Haber 2005 explains that:

“The impetus for industrial development came from the expansion of foreign trade. Driving the growth of foreign trade were two factors. The first was that most Latin American countries were on the silver standard, and silver fell in value relative to gold in the last two decades of the nineteenth century. Most Latin American countries therefore saw their currencies depreciate in real terms relative to the gold-backed currencies of the economies of the North Atlantic. As international trade theory would predict, real exchange rate depreciation resulted in the expansion of the tradables (e.g. industrial goods) [...] Second, the late nineteenth century also saw a dramatic decline in the international costs of transport, as steel-hulled steamships came to replace wood and sail.”

From Conflict to Cooperation

Why do lower levels of **sectoral inequality** (which implied **higher military threats**) lead to **sectoral cooperation**?

The rising of the industrial sector allowed industrial political elites to get access to military capacities that were as good as the agricultural elite's. The **threat** is what leads to **cooperation** rather than **conflict**. It makes no sense to engage in conflict when (1) both groups have the same 'fire power' and (2) when there is a cheaper exit (sectoral bargains).

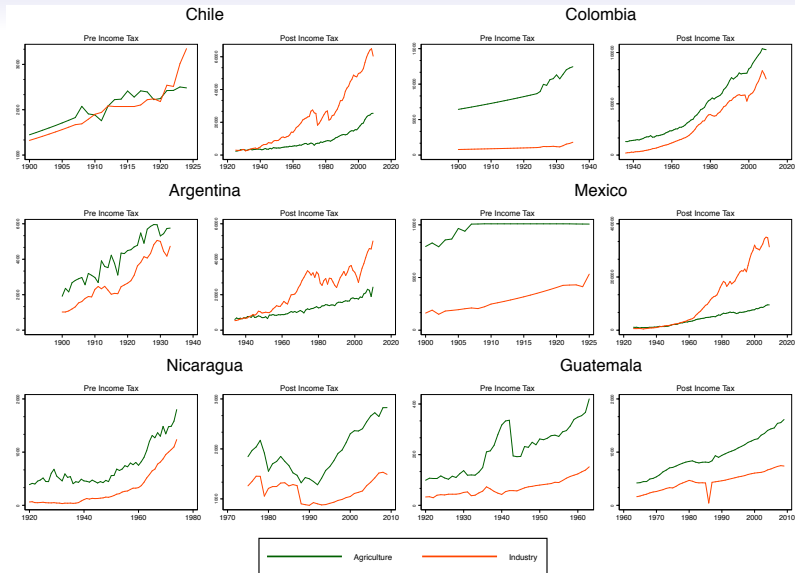
War was in 1891, but Income Tax in 1924?

- Civil wars of 1851-859 and 1891 between a “large landed property [elite against a] productive capital [elite].”
- President **Balmaceda's overthrowing in 1891** explains the sectoral nature of these conflicts.
- He was mainly supported by the landed elites, but later overthrown in 1891 by a mainly industrial/mining coalition:
 - His agenda on “industrial” infrastructure benefited mostly agricultural areas.
 - his attitude towards the banking sector (closely linked to the mining sector) confiscatory.
- At the same time, however, he failed to secure a coalition with his own sector.
 - Decline of wheat exports. Balmaceda's policies fostered sectoral dependence of agriculture on industrial production, forcing the “landed proprietors [to] become dependent to a considerable extent on the continuing prosperity of the major nitrate capitalists.” (Zeitlin).
 - While it would be inaccurate to say that Balmaceda was *completely* supported by agriculturalists and *completely* opposed by industrialists, this example illustrates how (failed) inter-sectoral alliances and biased public goods provision against industrialists led these two groups to a military conflict in 1891.
 - The conflict left a permanent scar in the Chilean society. While the civil war lasted only nine months, it took 10,000 lives (out of a total population of 3 million people) and cost more than \$ 100 million, a significant amount for a small country.
 - There was an intention to avoid more violence. For instance, while all “ministers, counselors of state, members of the constituent congress [,] municipal officials, provincial governors and intendants, members of the judiciary and even the lowest functionaries and ordinary employees of Balmaceda's government were investigated [or] brought to trial,” there were a number of amnesties issued. Similarly, there were a number of *aborted* coups in 1907, 1912, 1915 and 1919. I identify a third additional factor. War was more likely to exhaust all existent assets without producing positive outcomes for either sector, putting pressures for a sectoral compromise.

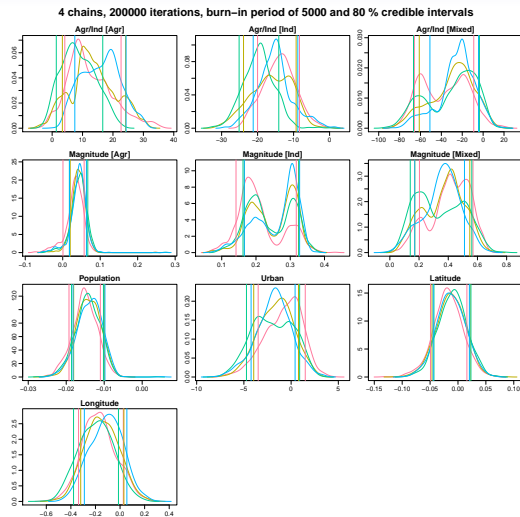
| Country | Time Frame | Sector | Augmented Dickey-Fuller | Phillips-Perron | KPSS | Conclusion |
|-----------|------------|-------------|-------------------------|-----------------|--------------------|------------|
| Chile | Pre | Agriculture | -1.185 (0.68) | -1.241 (0.66) | .107 [†] | I(1) |
| | | Industry | 2.310 (0.99) | 2.936 (0.99) | .115 [†] | I(1) |
| | Post | Agriculture | 4.957 (1.00) | 5.40 (1.00) | .289 | I(1) |
| | | Industry | 0.908 (0.99) | 1.458 (0.99) | .249 | I(1) |
| | All | Agriculture | 5.521 (1.00) | 6.722 (1.00) | .31 | I(1) |
| | | Industry | 1.582 (0.99) | 2.305 (0.99) | .314 | I(1) |
| Colombia | Pre | Agriculture | 2.709 (0.99) | 2.414 (0.99) | .204 | I(1) |
| | | Industry | 2.103 (0.99) | 3.297 (1.00) | .183 | I(1) |
| | Post | Agriculture | 2.382 (0.99) | 3.196 (1.00) | .282 | I(1) |
| | | Industry | 0.520 (0.98) | 1.044 (0.99) | .241 | I(1) |
| | All | Agriculture | 4.296 (1.00) | 5.893 (1.00) | .372 | I(1) |
| | | Industry | 1.674 (0.99) | 2.707 (0.99) | .374 | I(1) |
| Argentina | Pre | Agriculture | -0.849 (0.80) | -1.201 (0.67) | .0801 [†] | I(1) |
| | | Industry | -0.495 (0.89) | -0.378 (0.91) | .115 [†] | I(1) |
| | Post | Agriculture | 1.197 (0.99) | 1.093 (0.99) | .277 | I(1) |
| | | Industry | 0.228 (0.97) | 0.381 (0.98) | .0901 [†] | I(1) |
| | All | Agriculture | 1.484 (0.99) | 1.401 (0.99) | .332 | I(1) |
| | | Industry | 1.007 (0.99) | 1.237 (0.99) | .183 | I(1) |
| Mexico | Pre | Agriculture | 4.601 (1.00) | 5.952 (1.00) | .288 | I(1) |
| | | Industry | 5.803 (1.00) | 10.776 (1.00) | .29 | I(1) |
| | Post | Agriculture | 0.999 (0.9879) | 0.497 (0.99) | .100 [†] | I(1) |
| | | Industry | -1.295 (0.65) | -0.982 (0.76) | .113 [†] | I(1) |
| | All | Agriculture | 3.431 (1.00) | 3.607 (1.00) | .341 | I(1) |
| | | Industry | 0.672 (0.99) | 2.020 (0.99) | .367 | I(1) |
| Nicaragua | Pre | Agriculture | 2.473 (0.99) | 2.395 (0.99) | .25 | I(1) |
| | | Industry | 4.958 (1.00) | 9.100 (1.00) | .244 | I(1) |
| | Post | Agriculture | -0.154 (0.94) | 0.194 (0.97) | .2 | I(1) |
| | | Industry | -1.237 (0.6577) | -1.176 (0.68) | .189 | I(1) |
| | All | Agriculture | 0.636 (0.99) | 0.739 (0.99) | .116 [†] | I(1) |
| | | Industry | -0.164 (0.94) | -0.060 (0.95) | .123 | I(1) |
| Guatemala | Pre | Agriculture | -0.393 (0.91) | -0.343 (0.92) | .0639 [†] | I(1) |
| | | Industry | 1.358 (0.99) | 1.704 (0.99) | .199 | I(1) |
| | Post | Agriculture | 1.786 (0.99) | 1.905 (0.99) | .162 | I(1) |
| | | Industry | -0.998 (0.75) | -1.352 (0.61) | .0919 [†] | I(1) |
| | All | Agriculture | 3.349 (1.00) | 3.714 (1.00) | .321 | I(1) |
| | | Industry | 0.413 (0.98) | 0.017 (0.96) | .288 | I(1) |

| Country | Number of Cointegrated Vectors (rank) | Restrictions | Lags | Log-Likelihood | Trace |
|-----------|---------------------------------------|---------------------|------|----------------|---------|
| Chile | at least 1 | Restricted Constant | 5 | -1665.9736 | 0.3799 |
| Argentina | at least 1 | Restricted Constant | 3 | -1802.292 | 4.7657 |
| Colombia | at least 1 | Restricted Trend | 2 | -1805.6773 | 10.0076 |
| Mexico | at least 1 | Restricted Constant | 4 | -1978.1322 | 1.0274 |
| Nicaragua | 0 | Restricted Constant | 2 | -1020.221 | 11.5297 |
| Guatemala | 0 | Trend | 3 | -859.2802 | 16.5493 |

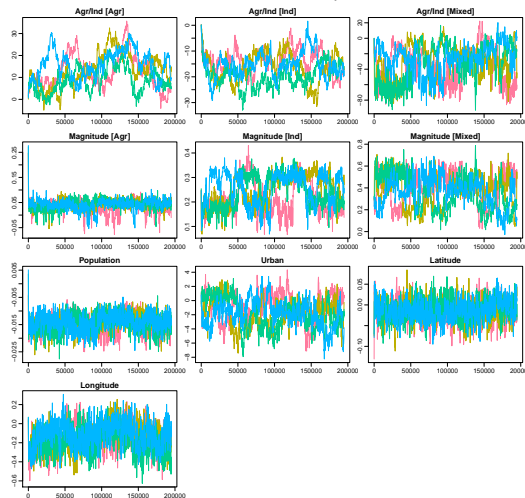
| Country | Time Frame | Number of Lags | LM | Normally Tests | | | Stability Condition |
|-----------|------------|----------------|----|----------------|----------|----------|---------------------|
| | | | | Jarque-Bera | Skewness | Kurtosis | |
| Chile | Pre | 4 | ✓ | ✓ | ✓ | ✓ | ✓ |
| | Post | 2 | ✓ | ✓- | ✓- | ✓- | ✓ |
| Colombia | Pre | 1 | ✓- | ✗ | ✗ | ✗ | ✓ |
| | Post | 1 | ✓ | ✓- | ✓- | ✓- | ✓ |
| Argentina | Pre | 2 | ✓ | ✓ | ✓ | ✓ | ✓ |
| | Post | 2 | ✓ | ✓- | ✓ | ✓- | ✓ |
| Mexico | Pre | 1 | ✓ | ✓- | ✓- | ✓- | ✓ |
| | Post | 2 | ✓ | ✓ | ✓ | ✓ | ✓ |
| Nicaragua | Pre | 2 | ✓ | ✓- | ✓- | ✓- | ✓ |
| | Post | 1 | ✓ | ✓- | ✓- | ✓- | ✓ |
| Guatemala | Pre | 3 | ✓ | ✗ | ✓- | ✓- | ✓ |
| | Post | 1 | ✓- | ✓- | ✓- | ✓- | ✓ |

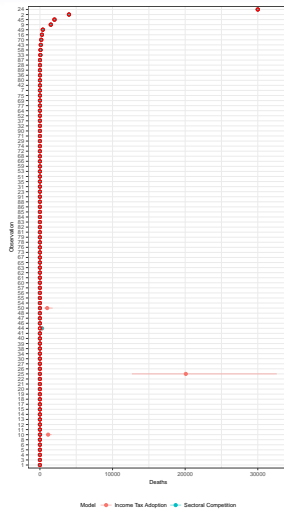


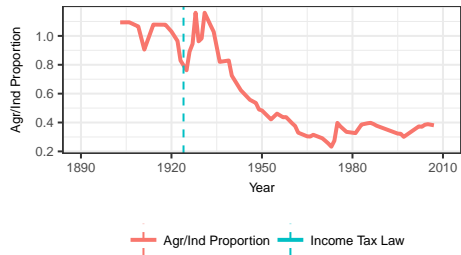
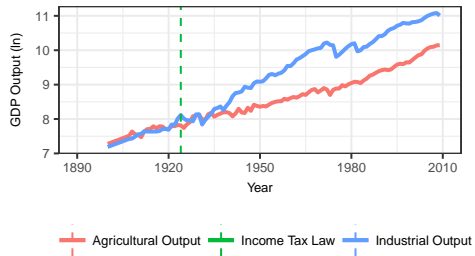
| Country | Pre/Post Income Tax | Sample | Directionality | chi2 | P-value |
|-----------|---------------------|-------------|------------------------|-------|---------|
| Chile | Pre | 1905 - 1924 | Agriculture → Industry | 3.55 | 0.47 |
| | | | Industry → Agriculture | 12.13 | 0.02 |
| | Post | 1928 - 2009 | Agriculture → Industry | 11.92 | 0.00 |
| | | | Industry → Agriculture | 5.37 | 0.07 |
| Colombia | Pre | 1902 - 1935 | Agriculture → Industry | 4.96 | 0.03 |
| | | | Industry → Agriculture | 10.44 | 0.00 |
| | Post | 1938 - 2009 | Agriculture → Industry | 4.32 | 0.04 |
| | | | Industry → Agriculture | 1.63 | 0.20 |
| Argentina | Pre | 1903 - 1933 | Agriculture → Industry | 4.19 | 0.12 |
| | | | Industry → Agriculture | .42 | 0.81 |
| | Post | 1937 - 2010 | Agriculture → Industry | .18 | 0.91 |
| | | | Industry → Agriculture | 1.37 | 0.50 |
| Mexico | Pre | 1902 - 1965 | Agriculture → Industry | .73 | 0.39 |
| | | | Industry → Agriculture | 11.57 | 0.00 |
| | Post | 1969 - 2009 | Agriculture → Industry | 5.56 | 0.06 |
| | | | Industry → Agriculture | 1.32 | 0.52 |
| Nicaragua | Pre | 1923 - 1974 | Agriculture → Industry | .48 | 0.79 |
| | | | Industry → Agriculture | 6.83 | 0.03 |
| | Post | 1977 - 2009 | Agriculture → Industry | .014 | 0.91 |
| | | | Industry → Agriculture | 4.96 | 0.03 |
| Guatemala | Pre | 1924 - 1963 | Agriculture → Industry | 2.18 | 0.54 |
| | | | Industry → Agriculture | 6.72 | 0.08 |
| | Post | 1966 - 2009 | Agriculture → Industry | .58 | 0.45 |
| | | | Industry → Agriculture | 6.05 | 0.01 |



4 chains, 200000 iterations and burn-in period of 5000







| | Mean | SD | Lower | Upper | Pr. |
|-------------------|--------|-------|--------|-------|------|
| Agr/Ind [Agr] | 12.68 | 7.21 | 3.73 | 22.65 | 0.98 |
| Agr/Ind [Ind] | -16.26 | 5.30 | -23.17 | -9.62 | 1.00 |
| Agr/Ind [Mixed] | -30.73 | 21.74 | -63.78 | -4.89 | 0.95 |
| Magnitude [Agr] | 0.04 | 0.02 | 0.01 | 0.06 | 0.95 |
| Magnitude [Ind] | 0.24 | 0.07 | 0.16 | 0.32 | 1.00 |
| Magnitude [Mixed] | 0.37 | 0.14 | 0.17 | 0.55 | 1.00 |
| Latitude | -0.01 | 0.03 | -0.05 | 0.02 | 0.69 |
| Longitude | -0.16 | 0.14 | -0.34 | 0.03 | 0.85 |
| Population | -0.01 | 0.00 | -0.02 | -0.01 | 1.00 |
| Urban | -1.54 | 2.01 | -4.22 | 1.00 | 0.76 |

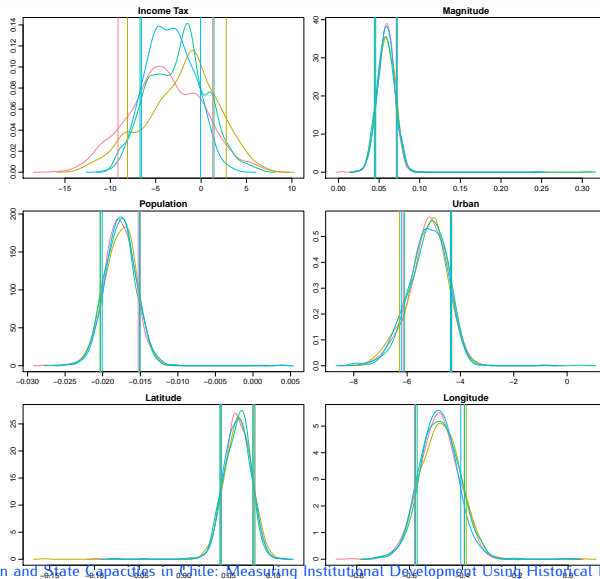
Note: 200000 iterations with a burn-in period of $n = 5000$ iterations discarded.

80% credible intervals (upper/lower bounds). All R-Hat statistics below critical levels.

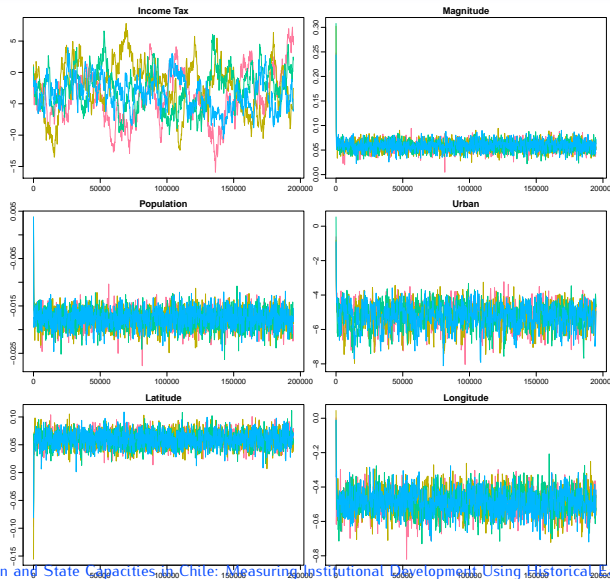
Standard convergence diagnostics suggest good mixing and convergence.

Year fixed effects were omitted in the table.

4 chains, 200000 iterations, burn-in period of 5000 and 80 % credible intervals



4 chains, 200000 iterations and burn-in period of 5000



| | Mean | SD | Lower | Upper | Pr. |
|------------|-------|------|-------|-------|------|
| Income Tax | -3.01 | 3.55 | -7.55 | 1.41 | 0.81 |
| Magnitude | 0.06 | 0.01 | 0.04 | 0.07 | 1.00 |
| Latitude | 0.06 | 0.01 | 0.04 | 0.08 | 1.00 |
| Longitude | -0.49 | 0.07 | -0.58 | -0.39 | 1.00 |
| Population | -0.02 | 0.00 | -0.02 | -0.02 | 1.00 |
| Urban | -5.22 | 0.73 | -6.19 | -4.35 | 1.00 |

Note: 200000 iterations with a burn-in period of $n = 5000$ iterations discarded.

80% credible intervals (upper/lower bounds). All R-Hat statistics below critical levels.

Standard convergence diagnostics suggest good mixing and convergence.

Year fixed effects were omitted in the table.

A total of 4 chains were run. Detailed diagnostic plots available [here](#).

For example, the historian Barros (1970) explains that before the civil war, *salitreras* (nitrate towns) in northern Chile were locally so important that they were considered “a state within the state.” **Local bosses had to approve decisions on whether public employees could be fired, whether public works could be developed, and on whether politicians could give public speeches.** Moreover, **they coined their own currency** and had their own particular **local laws**.