

# Internal Trade in India: Patterns and Determinants

ISI Annual Conference on Economic Growth and Development 2021

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December 21, 2021

- Vast scholarship on India's foreign trade patterns but sporadic research on internal trade flows
- Poor data availability for internal trade flows:
  - No publicly available data for goods transported by roads
  - Highly aggregated
- Limited understanding of internal trade drivers despite large variations in geographic units that trade with each other
- Trade flows intimately linked with trade barriers (Van Leemput (2021)) and costs (Abeberese and Chen (2021)) → important implications for local economic development

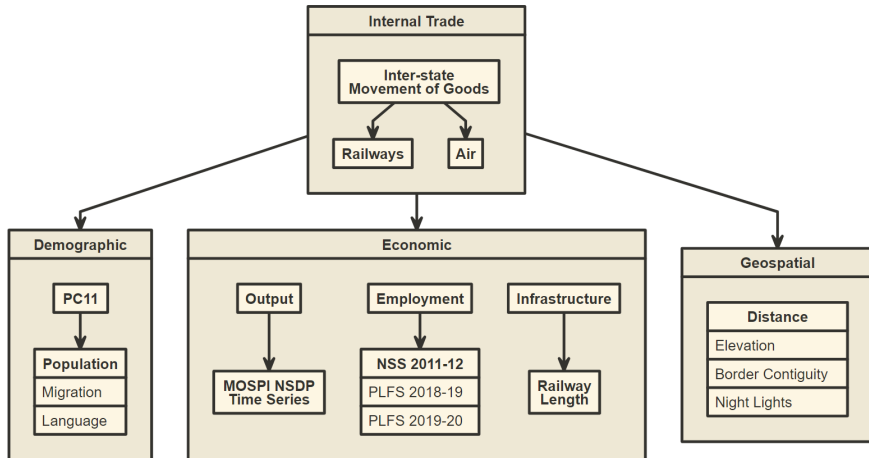
# Context: Rising Trade within India

- India handles  $\sim 4.6$  billion tonnes of goods each year  
→ total value of 950k cr.
- Policy push to improve logistics performance:
  - Development of Dedicated Freight Corridor
  - Build-out of road infrastructure projects → Bharatmala, national highway expansion
  - Capacity augmentation of waterways: Sagar Mala, Jal Vikas Marg Project



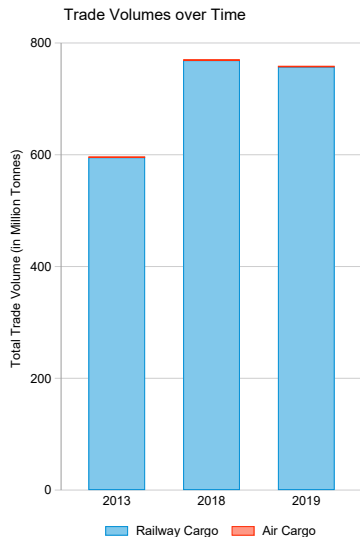
# Preview of Findings

- Internal trade within India's borders is becoming **spatially concentrated**  
→ fewer states are trading more
- Product basket also becoming concentrated over time: Bilateral trade flows dominated by few commodities
- **Product concentration matters:** Key determinant of how much states trade with each other



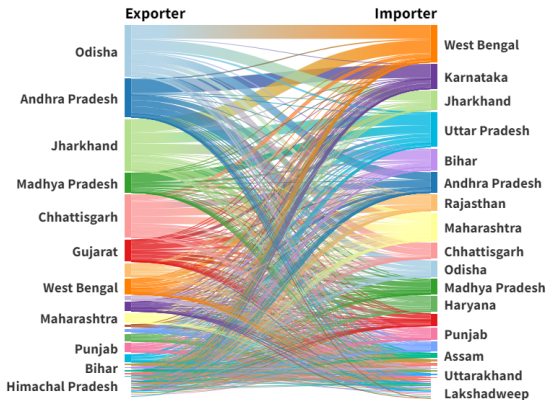
# Overall Trends

- Railway → dominant mode of transport
- Total trade volume: 600M tonnes in 2013-14, 770M tonnes in 2018-19 (28.3% increase in 5 years)
- 760M tonnes in 2019-20 (1.3% decline in 1 year)



# Flow of Goods by Railways and Air

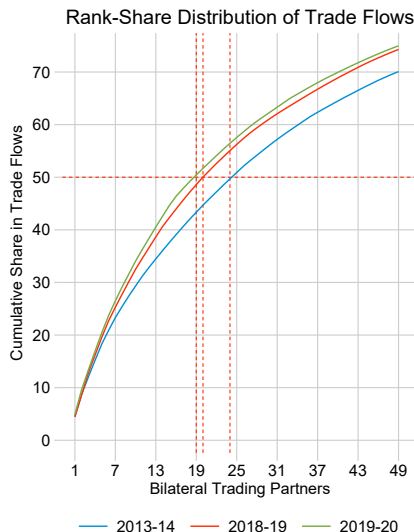
## Domestic Trade by Railways and Air (2018)



Source: DGCIS • Figures in million tonnes

**Figure:** Sankey diagram depicting flow of goods by Railways and Air in 2018

# Spatial Concentration



- Fewer states are trading more
- 2013: 24 trading partners accounted for 50% of all trade flows → down to 20 partners in 2018-19; 19 in 2019-20

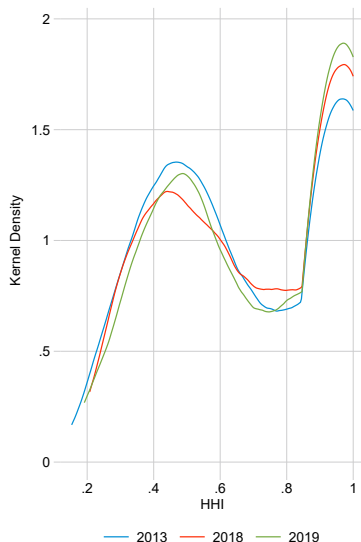


# Product Concentration

- Measure of product concentration:  
**Herfindahl-Hirschman Index**

$$H_{ij} = \sum_{k=1}^K (s_{ij})_k^2 \in [0, 1]$$

- Leverage ITCHS codes to construct H-Index measure for each bilateral trading partner
- $\overline{H}_{ij}$  in 2013: 0.24;  $\overline{H}_{ij}$  in 2018: 0.26;  $\overline{H}_{ij}$  in 2019: 0.27
- Top commodities traded: **Iron ore** (OR  $\rightarrow$  WB), **Coal** (JH  $\rightarrow$  WB), **Cement** (MP  $\rightarrow$  UP)



# Determinants of Trade: Gravity Model

- Structural gravity model in a partial equilibrium framework
- Consumer preferences represented by a CES utility function ((Novy (2013), Behrens et al. (2014), Arkolakis et al. (2012))

## Comparison

Standard Model	Structural Model
$Y_{ij} = C \frac{X_i X_j}{t_{ij}}$	$Y_{ij} = \frac{X_i E_j}{X} \cdot \frac{1}{d_{ij}}$

where,

- $Y_{ij}$  = trade flows from exporter state  $i$  to destination state  $j$
- $t_{ij}$  = trade cost term (barriers);  $d_{ij}$  = Distance between state  $i$  and  $j$
- $X_i$  and  $E_j$  = sizes (value of output) of state  $i$  and state  $j$  respectively

- We augment the reduced-form gravity model with a panel structure and additional covariates

$$\underbrace{Y_{ijt}}_{\text{Total trade volume in year } t} = \alpha_i + \underbrace{X_{it}\beta_1}_{\text{Output of Exporter in year } t} + \underbrace{E_{jt}\beta_2}_{\text{Output of Importer in year } t} + \underbrace{Z'_{ijt}\gamma}_{\text{Time-varying covariates}} + \underbrace{B'_{ij}\omega}_{\text{Time-invariant covariates}} + c_{ij} + u_{ijt}$$

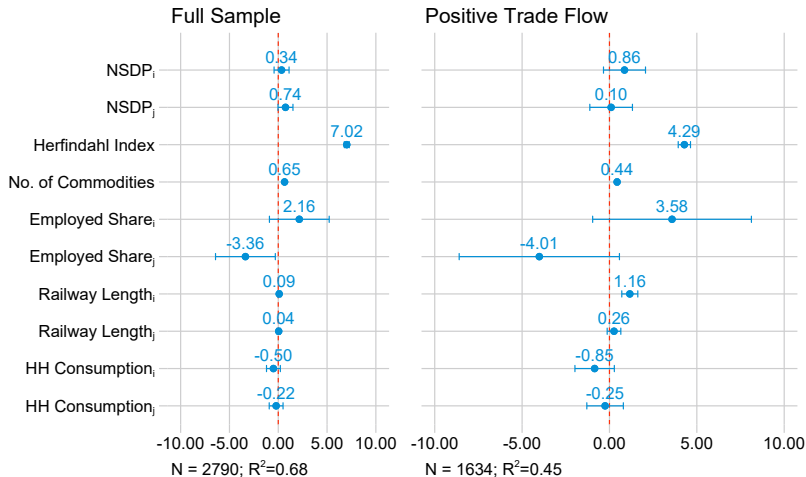
- To account for **zero-trade flows**: Inverse Hyperbolic Sine Transformation (Bellamare (2020))

$$\tilde{x} = \sinh^{-1}(x) = \ln \left( x + \sqrt{x^2 + 1} \right) \quad (1)$$

## Model Selection:

- Breusch-Pagan LM Test:  $P > \chi^2 = 0 \rightarrow \text{RE} > \text{POLS}$  ▶ RE / POLS result
- Durbin-Wu-Hausman Test:  $P > \chi^2 = 0 \rightarrow \text{FE} > \text{RE} (> \text{POLS})$
- Use **Fixed Effects estimator** to estimate parameters of the gravity model

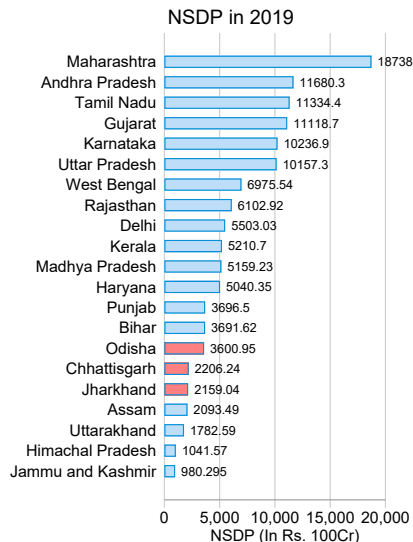
# Results: Fixed Effects Estimation



All values (except total commodities) are IHS transformed

# Mechanism: Resource Curse

- States that trade more by railways and air (CH, OR, JH) are not economically prosperous
- States rich in resources end up worse off economically (Dhillon (2016))
- Any gains from internal trade don't translate into economic growth

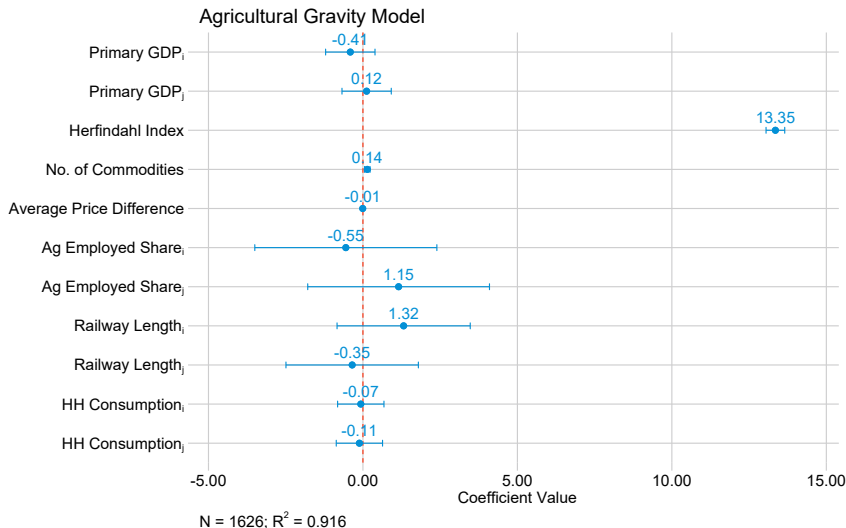


# From Volumes to Values: Agricultural Gravity Model

- Previous model only accounted for **volume**, not **value** of goods traded
- Agricultural goods:
  - Identifiable by ITCHS codes
  - Agricultural Price data: AgMarkNet database ([agmarknet.gov.in](http://agmarknet.gov.in))
  - Manually map ITCHS codes to AgMarkNet products
- Construct price difference measure a la Leemput (2021): average out daily modal prices, winsorize at 99th percentile; again average at year-level to get commodity prices



# Results: FE Agricultural Gravity Model



# Potential Explanations

- **Efficiency gains from infrastructural development:** Increased productivity → firms prioritize production of goods in which they are most efficient (Abeberese and Chen (2021))
- **Quality hypothesis:** States that specialize in the production of certain goods tend to export a higher average quality of the good (Alcala (2014)) → higher demand
- **Measurement Error:** Non-bulk goods are increasingly transported by roads due to lower trade costs
- **Declining trade barriers:** Lower bureaucratic complexity, harmonized rates under IGST



- Need for **inclusionary internal trade policy**: integrate remote geographies in internal trade
- Bigger push to transportation infrastructural development: Potential for increased efficiency gains and product specialization
- Improvements in **data ecosystem**:
  - Trade flow data for roads: E-Way Bills
  - Spatially and temporally disaggregated data: **Trade is a highly local phenomenon**

**Thank you! Questions?**

# Results: Pooled OLS v/s Random Effects

## ► Empirical Strategy

