# Investment in Infrastructure and Trade: The Case of Ports

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#### Introduction

- Transportation infrastructure is crucial for the smooth functioning of international trade
- Ports are vital links of the infrastructure network
  - > 80% of trade carried by ships
  - crucial determinants of trade costs
- This paper:
  - Port production function
    - Queueing theory; capture congestion dynamics
    - Input: infrastructure, labor, cranes; output: time
    - Congestion highly sensitive to demand shocks
  - Demand system for transportation services
    - Exporter-importer pairs decide which port to choose
    - Factors: time, distance, port prices, other port characteristics
    - Macroeconomic conditions shift demand over time
    - Agents sensitive to changes in time at port
  - Counterfactuals: welfare gains of port infrastructure investment
    - One additional ship for one port

#### Introduction

# Paper Findings

- Port Investment → substantial trade and welfare gains only if targeted properly
- Sizeable spillovers across ports
- Macroeconomic volatility  $\rightarrow$  returns to investment & geography

#### Literature

- transportation infrastructure on trade (e.g. Redding and Turner, 2015, Donaldson and Hornbeck, 2016, Redding,
   2016, Donaldson, 2012, Fajgelbaum and Schaal, 2020, Allen and Arkolakis, 2022)
- impacts of ports; shipping industry
- congestion in the context of urban transportation (e.g. Durrmeyer and Martinez, 2022, Bordeu, 2023, Kreindler, 2023, Almagro et al., 2024)
- supply chain disruptions
- international trade, the role of geography, and the trade cost (e.g. Eaton and Kortum, 2002, Anderson and Van Wincoop, 2003, Melitz, 2003)

# Background- Port Operation

#### - Procedures

- ship waits in anchorage  $\rightarrow$  enter the port to get serviced in berth  $\rightarrow$  cargo loaded by workers & cranes and stored  $\rightarrow$  cargo wait for trucks or trains to final destination  $\rightarrow$  ship departs

#### - Port Infrastructure

- Includes berth, cranes, storage areas, buildings, etc.
- Port services: pilotage, cargo handling, towage; 40 services per ship per visit
- Investment: \$45 bi in 2012-2016; \$163 bi scheduled in 2021-2025
   mix of private & local public funds in the US; serve local interests (e.g., high throughput, employment, and connectivity)

## Focus of This Paper

- Dry bulk carriers- commodities and raw materials such as iron ore, steel, coal and grain;
   1/2 total seaborne trade worldwide
- 51 largest heterogeneous US ports

#### Data

- Port Production Function
  - Universe of port calls:

AXS Marine (AXS Dry) Data for bulk carriers > 10,000 DWT

Vars: timestamp for vessel's arrival/ entry to and exit form the port, an indicator for whether the ship loaded/discharged, estimate for the commodity on board

- Port Infrastructure:
  - Miles of each berth, acres of storage space, and number of cranes from Google Maps and Google Earth Pro
- Port Employment:
   Longshoremen union membership form the Department of Labor
- Port Demand
  - Monthly number of arrivals at each of the ports
     AXS Marine + Freight Analysis Framework Regional Database
  - Port charges for each service requested
     Anonymous provider; ~ 35% global port calls Vars: berth dues, cargo handling dues, line handling, pilotage
- Port Infrastructure Expansion Cost
  - Dredging: site-specific cost estimates for dredging a new berth using cost projections from US
     Army Corp of Engineers
  - Acquisition of Land for Storage: land value estimates from Nolte, 2000 and industrial & commercial land data from OpenStreetMan

#### Data

# Summary Statistics

- Port calls (throughput in tons): total annual number of ships (total tonnage) that were handled at

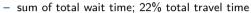
	Mean	$^{\mathrm{SD}}$	25th percentile	75th percentile
Annual number of port calls	136	130	54	180
Annual throughput (tons)	7,295,140	7,355,597	2,602,124	9,970,174
Time at port (hours): average	117	54	74	153
Time at port (hours): st. dev. over time	88	41	58	118
Wait time (hours): average	39	39	2.9	58
Wait time (hours): st. dev. over time	58	50	16	101
Fraction of port calls with positive wait time	41%	30%	13%	70%
Port dues (\$ per ton)	2.1	1.3	1.2	2.9
Port dues (in \$) per port call	117,345	96,206	44,038	168,399
Length of berths (miles)	0.96	0.8	0.42	1.4
Storage space (acres)	176	176	55	234

#### port

- Wait/Congestion time more volatile than service time
- Ports do NOT specialize in specific commodities

# Port Efficiency

## - Time at Port





- M/M/K queueing models:
  - ship arrivals follow a Poisson process and port service times are exponentially distributed
- Port Technology
  - Ship's expected total time at port:

$$\underbrace{T}_{\text{service time}} + \mathbb{I}\{Q \ge K\} \underbrace{(Q - K + 1)\frac{T}{K}}_{\text{wait time (queueing)}} \tag{1}$$

K: num of ships the port can handle at most at a time

T: expected service time at port

Q: (endogeneous) num of ships that the incoming ship finds ahead

$$T\left(\frac{L}{s}, \frac{c}{s}, \omega\right) \tag{2}$$

L: port's workers

s: num of ships currently serviced

c: cranes;  $\omega$ : productivity

## - Port Technology

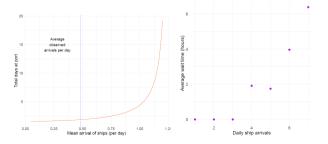
- Markov structure of endogenous evolution of the port queue
- Prob  $p_n( au)$  that in instant au there are n total ships at port, either being serviced or waiting

$$\frac{dp_{n}(\tau)}{d\tau} = \begin{cases} -\lambda p_{0}(\tau) + \mu(0) \, p_{1}(\tau), & \text{if } n = 0, \\ -\left(\lambda + n \, \mu(n)\right) \, p_{n}(\tau) + \lambda p_{n-1}(\tau) + (n+1)\mu(n)p_{n+1}(\tau), & \text{if } 0 < n < K, \\ -\left(\lambda + K \, \mu(K)\right) \, p_{n}(\tau) + \lambda p_{n-1}(\tau) + K \, \mu(K)p_{n+1}(\tau), & \text{if } n \ge K. \end{cases}$$

 $\lambda$ : ship arrival rate

 $\mu(s) = 1/T(s)$ : rate at which ship completes service

Convex Cost of Congestion (Simulation)



#### Port Demand

- Payoff to shipment i between d and f from choosing port j in month t:

$$u_{ijt} = \beta_f \operatorname{dist}(f, j) + \beta_d \operatorname{dist}(j, d) - \beta_T T T_{jt} - \beta_p p_{jt} + \gamma_t + \gamma_f + \gamma_{l(j)} + \xi_{jft} + \epsilon_{ijt}$$
 (4)

 $\begin{array}{l} \operatorname{dist}(f,j) \colon \operatorname{distance} \ \operatorname{between} \ \operatorname{the} \ \operatorname{foreign} \ \operatorname{location} \ f \ \operatorname{and} \ \operatorname{port} \ j \\ TT_{jt} \colon \ (\operatorname{endogenous}) \ \operatorname{total} \ \operatorname{time} \ \operatorname{at} \ \operatorname{port} \ j \ \operatorname{in} \ \operatorname{period} \ t \\ p_{jt} \colon \operatorname{price} \ \operatorname{that} \ \operatorname{port} \ j \ \operatorname{charges} \ \operatorname{for} \ \operatorname{its} \ \operatorname{services} \ \operatorname{in} \ \operatorname{month} \ \operatorname{t} \\ \gamma_t \colon \ \operatorname{month} \ \operatorname{fixed} \ \operatorname{effects} - \ \operatorname{macro} \ \operatorname{economic} \ \operatorname{fluctuations} \\ \gamma_f \ \& \ \gamma_{l(j)} \colon \operatorname{origin} \ \& \ \operatorname{destination} \ \operatorname{fixed} \ \operatorname{effects} \\ \xi_{jft} \colon \ \operatorname{unobserved} \ \operatorname{demand} \ \operatorname{shocks} \ \operatorname{for} \ \operatorname{period} \ t \\ \epsilon_{ijt} \colon \operatorname{iid} \ \operatorname{shipment-specific} \ \operatorname{shock}; \ \operatorname{Type} \ \operatorname{IEV} \ \operatorname{distribution} \end{array}$ 

- Advantages measure the welfare cost of delays-  $\beta_T T T_{jt}$  substitutions patterns depend on foreign and domestic locations

## - Macroeconomic Volatility

Macroeconomic conditions affect overall demand for ports:

$$\gamma_t = \rho_0 + \rho_1 \gamma_{t-1} + \epsilon_t \tag{5}$$

$$\epsilon_t \sim N(0, \sigma_\epsilon^2)$$

#### Trade Flows

- Probability that a pair trading between foreign location f and domestic location d chooses port f:

$$\sigma_{jfdt} \equiv \Pr(j \mid f, d, t) = \frac{\exp(\delta_{jft} + \beta_d \operatorname{dist}(j, d))}{1 + \sum_l \exp(\delta_{lft} + \beta_d \operatorname{dist}(l, d))}$$
(6)

$$\delta_{jft} \equiv \beta_f \operatorname{dist}(f, j) - \beta_T T T_{jt} - \beta_p p_{jt} + \gamma_t + \gamma_f + \gamma_{l(j)} + \xi_{jft}$$
(7)

- Problem:  $\sigma_{ifdt}$  not directly observable, use:

$$\sigma_{jft} \equiv \underbrace{\Pr(j \mid f, t)}_{\text{observed ship flows}} = \sum_{d} \Pr(j \mid f, d, t) * \underbrace{\Pr(d \mid f, t)}_{\text{observed US regional trade flows}}$$
 (8)

Monthly total trade/ship flow through port j:

$$\lambda_{jt} = \sum_{f} \sum_{d} \sigma_{jfdt} M_{df} \tag{9}$$

assume a ship arrives at port j at Poisson rate  $\lambda_{it}$ 

# Estimation Strategy- Production

- Port Technology
  - Service Time- CES

$$T_{jt} = \omega_{jt} \left( \alpha \left( \frac{L_{jt}}{s_{jt}} \right)^{\eta} + (1 - \alpha) \left( \frac{c_{jt}}{s_{jt}} \right)^{\eta} \right)^{-\frac{1}{\eta}}$$
 (10)

- Endogeneity: inputs are not orthogonal to realized productivity
  - IVs: 1. labor- local wages in related occupations
    - 2. changes to the labor force and to % the population above 65
    - 3. lagged inputs

GMM estimation:  $\mathbb{E}(\nu_{jt}z_{jt})=0$  , where  $\log \omega_{jt}=\log \omega_{j}+\delta t+\nu_{jt}$ 

Port Capacity- Leontief

$$K_j = \min\left\{\frac{B_j}{\kappa_1}, \frac{A_j}{\kappa_2}\right\} \tag{11}$$

 $B_j$ : port j's total length of its berths

 $A_j$ : acreage of storage space

 $\kappa_1$ : median number of miles of berth per ship  $(B_j/K_j)$ 

 $\kappa_2$ : median storage acreage per ship  $(A_j/K_j)$ 

## Estimation Result- Production

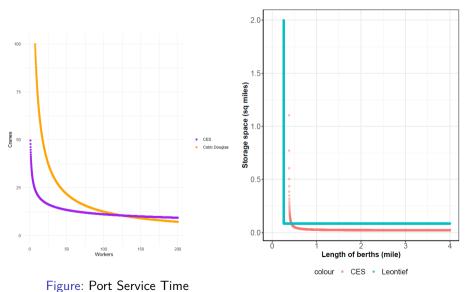


Figure: Port Capacity Function

# Estimation Strategy- Demand

#### Micro Likelihood

$$\mathcal{L}^{\text{micro}}(\delta, \beta_d) = \sum_{f, t} \sum_{j} n_{ft}^{j} \log \left( \Pr(j \mid f, t) \right)$$
(12)

$$= \sum_{f,t} \sum_{j} n_{ft}^{j} \log \left( \sum_{d} \Pr(j \mid f, d, t) \Pr(d \mid f, t) \right)$$
(13)

$$= \sum_{f,t} \sum_{j} n_{ft}^{j} \log \left( \sum_{d} \frac{\exp(\delta_{jft} + \beta_{d} \operatorname{dist}(j,d))}{1 + \sum_{l} \exp(\delta_{lft} + \beta_{d} \operatorname{dist}(l,d))} \operatorname{Pr}(d \mid f,t) \right)$$
(14)

- Problem: Endogeneity of  $TT_{jt}$ 

IVs: 1. unexpected disruptions in port j's service operations at t residuals from a regression of service time on a number of controls

Moment conditions: 
$$\mathbb{E}\left(\xi_{jft}z_{jft}\right)=0$$

Macro Likelihood

$$\mathcal{L}^{\mathsf{macro}}(\delta, \beta_d) = \sum_{j} \sum_{l} (\sigma_{fjd,\mathsf{year}})^{n_{fjd,\mathsf{year}}}$$
(15)

Purpose: identify  $\beta_d$   $n_{fid.vear}$ : annual flow from f to d through j

# Estimation Strategy- Demand

## - Objective Function

$$\arg\min_{\beta_{\delta}, \beta_{d}, \delta} \mathcal{L}^{\mathsf{micro}}(\delta, \beta_{d}) + \frac{1}{2} m(\delta, \beta_{\delta})' W m(\delta, \beta_{\delta}) + \alpha \mathcal{L}^{\mathsf{macro}}(\delta, \beta_{d})$$
 (16)

-  $m(\delta, \beta_{\delta})$ : empirical analog of  $\mathbb{E}\left(\xi_{jft}z_{jft}\right) = 0$ 

$$m(\delta, \beta_{\delta}) = \sum_{f, t} \sum_{j} z_{jft} \left( \delta_{jft} - \beta_f \operatorname{dist}(f, j) + \beta_T T T_{jt-1} + \beta_p p_{jt} - \gamma_t - \gamma_f - \gamma_{l(j)} \right)$$
(17)

Enrich with macroeconomic shocks

$$\gamma_t = \beta y_t + \epsilon_t \tag{18}$$

 $y_t$ : Baltic Dry Index, commodity prices, Industrial Price Index

# Estimation Result- Demand

Demand Estimates			
	1.44		
Average time at port	-1.44		
Don't and an area to a	(0.49)		
Port price per ton	-1.60		
T : 1: (1	(0.02)		
Foreign distance (days of travel)	-0.71		
Domestic distance (less days of travel)	$(0.15) \\ -1.2$		
Domestic distance (log days of travel)	<u>-</u>		
East Coast FE	(0.05) -0.98		
East Coast FE	(2.10)		
Great Lakes FE	-15.89		
Great Lakes FE	(4.57)		
Gulf FE	-8.22		
Gui I L	(3.73)		
	(5.75)		
Foreign location FE	Yes		
Month FE	Yes		
	200		
Instruments	Unexpected disruptions to port operations		

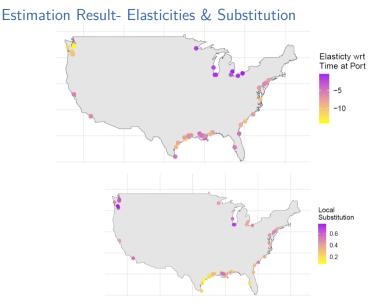


Figure 8: Local Substitution. For each port j this figure plots the fraction of shipments through j that is diverted away from j to a local port (i.e. to one of the 10 closest ports), conditional on shippers choosing to switch following a small increase in time at port.

# Welfare Gains from Infrastructure Investment (Counterfactuals)

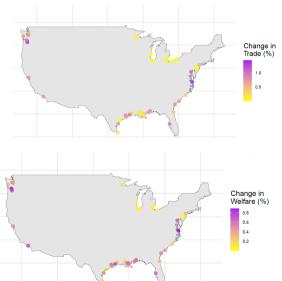
- Investment: one additional ship for each port  $K_j o K_j + 1$ 
  - assume j's labor and crane also increase
  - 30 years for this counterfactual
- Treated Ports
  - Congestion first declines due to higher capacity; then increases due to higher demand
  - On average (median) trade increases by 42% (38%) at the treated port; congestion declines on average (median) by 4.1% (3.7%)
- Other Ports
  - Substitution effects: lose demand
  - Feedback effect: de-congestion boost demand
  - On average trade decreases by 0.19%, while congestion declines on average by 0.6%
- Welfare Measure

$$\sum \beta^t W\left(TT_t, \gamma_t\right) \tag{19}$$

Where

$$W\left(TT_{t}, \gamma_{t}\right) = \frac{1}{\beta_{p}} \left| \sum_{d, f} \log \left( \sum_{j} \exp\left(\beta_{T}TT_{jt} + \gamma_{t} + x_{jfd}\right) + 1 \right) M_{df} + \gamma^{\text{euler}} \right| \tag{20}$$

Welfare Gains from Infrastructure Investment (Counterfactuals)



# Net Returns from Infrastructure Investment

Cost for Port:

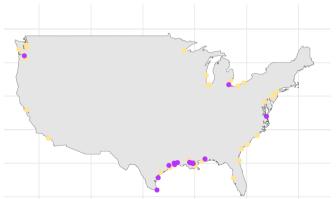
$$C_j^d + C_j^l + C_j^b \tag{21}$$

 $C_i^d$ : dredging cost

 $C_j^l$ : land purchasing cost

 $C_j^b$ : cost of constructing a bulkhead to support the new berth

- NetReturn = Welfare - Cost



# **Findings**

- Port Investment  $\rightarrow$  substantial trade and welfare gains only if targeted properly
- Sizeable spillovers across ports
- Macroeconomic volatility  $\rightarrow$  returns to investment & geography

Thank you!

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