

Worker Protectionism with Third Country Responses

Abdelrahman Amer
University of Toronto

Mahmood Haddara
University of Toronto

Daniel Trefler
University of Toronto

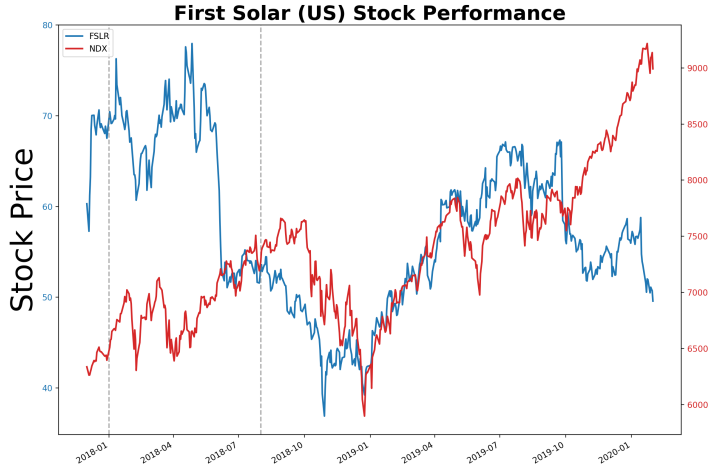
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Motivation

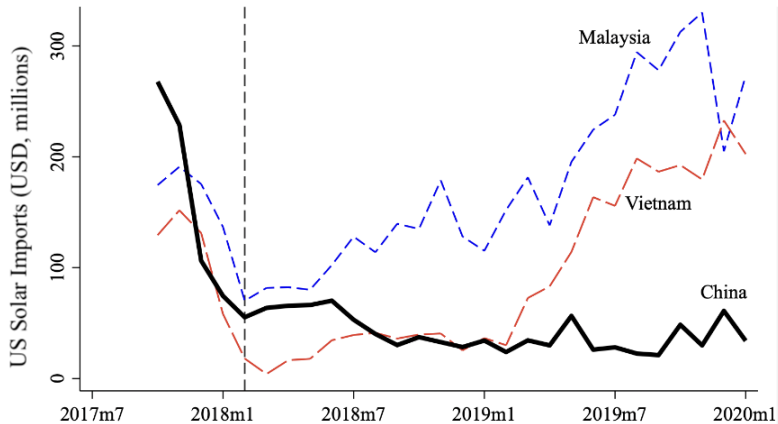
- It is generally accepted that tariffs involve a tradeoff between domestic jobs and aggregate prices
- To quantify this tradeoff, it is essential that models accurately capture production relocation
- Ex: If tariffs on F lead to production fully relocating to F' , there is no upside
- Standard models impose restrictive assumptions on the effect of $H \rightarrow F$ tariffs on F'
- This paper: Evaluate protectionist tariffs in a framework that accurately captures third country effects

- Third Country Effects: Data and Models
- Latent Factor Approach
- Model Description
- Solution Method

Third Country Effects: Data



Third Country Effects: Data



Third Country Effects: Data

- Flaaen et al. (2020): washing machines
 - Tariffs on Mexico and South Korea → Production relocates to China
 - Tariffs on China → Production relocates to Thailand and Vietnam
 - Production finally returns to the US after global tariffs... at a cost
 - The estimated annual consumer cost per job was \$815,000

Third Country Effects in EK

- Quantitative analyses of tariffs typically use the Eaton-Kortum (EK) framework
- Continuum of varieties, discrete choice for origin of each variety
- Like many discrete choice models, the EK framework exhibits **Independence of Irrelevant Alternatives (IIA)**

- Classic example from IO: Blue and Red Buses
- Consider a discrete choice between a blue bus and a train
- For simplicity assume that the current market share is 50-50
- How will market shares change if a new bus is introduced that is a different color (e.g. red)?
- Intuitive answer: 50% remains with trains, 50% split between red and blue bus
- Standard model (e.g. logit) answer: $x\%$ goes to red bus, $\frac{1-x}{2}\%$ goes to blue bus, $\frac{1-x}{2}\%$ goes to train
- Why? If the blue bus and train were equally attractive before, they should remain so afterwards

Third Country Effects in EK

Let π_{ni} denote the trade share of n in country i and consider a 3 country EK model:

$$\pi_{CH,US} = \frac{T_{CH}(x_{CH,US})^{-\theta}}{D}$$

$$\pi_{VN,US} = \frac{T_{VN}(x_{VN,US})^{-\theta}}{D}$$

$$\pi_{US,US} = \frac{T_{US}(x_{US,US})^{-\theta}}{D}$$

Where T_i is TFP, x_{ni} is the cost of buying from country n in country i , θ is the trade elasticity, and $D = \sum_{i \in US, CH, VN} T_i(x_{i,US})^{-\theta}$

Third Country Effects in EK

Consider a 10% increase in the US tariff on China

$$\hat{\pi}_{CH,US} = \hat{D}^{-1}(1.1)^{-\theta}$$

$$\hat{\pi}_{VN,US} = \hat{D}^{-1}$$

$$\hat{\pi}_{US,US} = \hat{D}^{-1}$$

Built into the EK model is the impossibility of Vietnam gaining more (proportionally) from China than the US (IIA).

- A well known departure from IIA is to apply discrete choice within “nests”
- Bus-train example: Outer nest (bus vs train) and inner nest (red vs blue)
- How to choose the nests in the context of trade?
- Previous models: Observables such as sector or geography
- Lind and Ramondo (AER, 2023): Latent Factor Estimation

Latent Factor Approach

- Key idea: What matters is the correlation of productivity draws between j_i and j'_i , more than observable labels
- Many factors are likely to govern this correlation, and not all of them are observable or easily modeled
 - Institutional knowledge
 - Brand penetration
 - Cultural awareness
 - Specific labor
 - ...
- Latent factor approach: Nested discrete choice, where nests are unobserved (latent)
- Allows for **product-level** heterogeneity in third country effects

Model

- Recall that tariffs are a tradeoff between higher wages (for some) and higher prices (for all)
- To distinguish between winners and losers, need heterogeneous households
- Follow Caliendo, Dvorkin, and Parro (ECMA, 2019): Households make a discrete choice over industry, with preference shocks and switching costs
 - $\Rightarrow$ heterogeneous wages, heterogeneous effects of tariffs

Households

- N countries and J industries
- Each country-industry pair (n, j) contains a representative household of endogenous size L_{nj}
- Log preferences over consumption of a final good C
- An employed household in (n, j) receives wage w_{nj} and purchases the final good at price P_n
- A nonemployed household, i.e. a household $(n, 0)$, receives consumption b^n

Industry Choice

- Households pay time invariant, additive utility costs $\tau_n^{jj'}$ to relocate from (j) to (j')
- In each period, households draw preference shocks for each industry $\epsilon_n^{j'}$
- The only decision households make in each period is where to relocate for next period
- In particular, the value of being in (n, j) at time t is

$$v_t^{nj} = \log(C_t^{nj}) + \max_{j'} \left\{ \beta \mathbb{E} [v_{t+1}^{nj'}] - \tau_n^{jj'} + \nu \epsilon_n^{j'} \right\} \quad (1)$$

Where $C_t^{nj} = b^n$ if $j = 0$ and $\frac{w_t^{nj}}{P_t^n}$ otherwise

- Assuming type I extreme value shocks with 0 mean, we get:

$$V_t^{nj} \equiv \mathbb{E} [v_t^{nj}] = \log (C_t^{nj}) + \nu \log \left(\sum_{j'} \exp \left(\beta V_{t+1}^{nj'} - \tau_n^{jj'} \right)^{\frac{1}{\nu}} \right) \quad (2)$$

- It follows that

$$\mu_t^{nj, nj'} = \frac{\exp \left(\beta V_{t+1}^{nj'} - \tau_n^{jj'} \right)^{\frac{1}{\nu}}}{\sum_k \exp \left(\beta V_{t+1}^{nk} - \tau_n^{jk} \right)^{\frac{1}{\nu}}} \quad (3)$$

- The distribution of labor evolves according to

$$L_{t+1}^{nj'} = \sum_j \mu_t^{nj, nj'} L_t^{nj} \quad (4)$$

- The final good in each i is produced according to

$$\left(\int_0^1 q_i(\varphi)^{\frac{\eta-1}{\eta}} d\varphi \right)^{\frac{\eta}{\eta-1}} \quad (5)$$

- For each φ , each (n, j) has a perfectly competitive firm that can produce φ and send it to i with linear productivity z_{nji} and observed tariffs t_{nji}
- Perfect competition implies

$$p_i(\varphi) = \min_{n,j} \left\{ \frac{t_{nji} w_{nj}}{z_{nji}(\varphi)} \right\} \quad (6)$$

- In each i , the vector of productivities for potential sources (n, j) is drawn iid according to

$$Pr[z_{11i}(\varphi) \leq z_{11}, \dots, z_{NJi}(\varphi) \leq z_{NJ}] = \exp \left[-G^i \left(T_{11i} z_{11i}^{-\theta}, \dots, T_{NJi} z_{NJi}^{-\theta} \right) \right] \quad (7)$$

with

$$G^i(x_{11}, \dots, x_{NJ}) = \sum_k \left[\left(\sum_n \sum_j (\omega_{knji} x_{nj})^{\frac{1}{1-\rho_k}} \right)^{1-\rho_k} \right] \quad (8)$$

- EK corresponds to $G(x_1, \dots, x_n) = \sum_j x_j$
- k indexes nests, ρ_k governs correlation within nest
- Every nji is in every nest, but weights differ

From Proposition 2 in Lind and Ramondo (2023), we have:

$$\sum_k \frac{a_{knji}}{\underbrace{\sum_{n'} \sum_{j'} a_{kn'j'i}}_{\text{Within}}} \frac{\left(\sum_{n'} \sum_{j'} a_{kn'j'i} \right)^{1-\rho_k}}{\underbrace{\sum_{k'} \left(\sum_{n'} \sum_{j'} a_{kn'j'i} \right)^{1-\rho_k}}_{\text{Between}}} \quad (9)$$

Where

$$a_{knji} = \left(\omega_{knji} T_{nji} w_{nj}^{-\theta} \right)^{\frac{1}{1-\rho_k}} \quad (10)$$

Solution Method

- Assume a fixed number of nests K
- Identification assumption: $T_{nji}\omega_{knji} = A_{kni}B_{jk}$
- $\{\rho_k, A_{kni}B_{jk}\} \rightarrow \text{unique } \{\pi_{nji}\}$
- Estimate parameters to minimize distance between model and data trade shares
- Continue increasing K until fit does not improve (LR stop at $K = 7$)

Solution Method

- Employ hat-algebra method as in Caliendo et al. (2019)
- No need to estimate switching costs τ
- Economy need not be in steady state: Simply input initial conditions+shocks and iterate
- Two step procedure: Outer loop over labor flows, inner loop over trade shares

$$\hat{w}_{nj} \hat{L}_{nj} w_{nj} L_{nj} = \sum_i \hat{\pi}_{nji} \pi_{nji} \left[\left(\sum_j \hat{w}_{ij} \hat{L}_{ij} w_{ij} L_{ij} \right) - TB_i \right] \quad (11)$$

$$\hat{\pi}_{nji} \pi_{nji} = \frac{\hat{T}_{nji} \hat{t}_{nji}^{-\theta} \hat{w}_{nj}^{-\theta} \tilde{\pi}_{nji} G_{nj}^i \left(\hat{T}_{11i} \hat{t}_{11i}^{-\theta} \hat{w}_{11}^{-\theta} \tilde{\pi}_{11i}, \dots, \hat{T}_{Nji} \hat{t}_{Nji}^{-\theta} \hat{w}_{NJ}^{-\theta} \tilde{\pi}_{Nji} \right)}{G^i \left(\hat{T}_{11i} \hat{t}_{11i}^{-\theta} \hat{w}_{11}^{-\theta} \tilde{\pi}_{11i}, \dots, \hat{T}_{Nji} \hat{t}_{Nji}^{-\theta} \hat{w}_{NJ}^{-\theta} \tilde{\pi}_{Nji} \right)} \quad (12)$$

$$\hat{P}_i = G^i \left(\hat{T}_{11i} \hat{t}_{11i}^{-\theta} \hat{w}_{11}^{-\theta} \tilde{\pi}_{11i}, \dots, \hat{T}_{Nji} \hat{t}_{Nji}^{-\theta} \hat{w}_{NJ}^{-\theta} \tilde{\pi}_{Nji} \right)^{-\frac{1}{\theta}} \quad (13)$$

$$\hat{w}_{nj} \hat{L}_{nj} w_{nj} L_{nj} = \sum_i \hat{\pi}_{nji} \pi_{nji} \left[\left(\sum_j \hat{w}_{ij} \hat{L}_{ij} w_{ij} L_{ij} \right) - TB_i \right] \quad (14)$$

$$\hat{\pi}_{nji} \pi_{nji} = \frac{\hat{T}_{nji} \hat{t}_{nji}^{-\theta} \hat{w}_{nj}^{-\theta} \tilde{\pi}_{nji} G_{nj}^i \left(\hat{T}_{11i} \hat{t}_{11i}^{-\theta} \hat{w}_{11}^{-\theta} \tilde{\pi}_{11i}, \dots, \hat{T}_{Nji} \hat{t}_{Nji}^{-\theta} \hat{w}_{NJ}^{-\theta} \tilde{\pi}_{Nji} \right)}{G^i \left(\hat{T}_{11i} \hat{t}_{11i}^{-\theta} \hat{w}_{11}^{-\theta} \tilde{\pi}_{11i}, \dots, \hat{T}_{Nji} \hat{t}_{Nji}^{-\theta} \hat{w}_{NJ}^{-\theta} \tilde{\pi}_{Nji} \right)} \quad (15)$$

$$\hat{P}_i = G^i \left(\hat{T}_{11i} \hat{t}_{11i}^{-\theta} \hat{w}_{11}^{-\theta} \tilde{\pi}_{11i}, \dots, \hat{T}_{Nji} \hat{t}_{Nji}^{-\theta} \hat{w}_{NJ}^{-\theta} \tilde{\pi}_{Nji} \right)^{-\frac{1}{\theta}} \quad (16)$$

Input $w, L, \pi, G, \theta, TB, \hat{T}, \hat{t}, \hat{L}$; Output $\hat{w}, \hat{\pi}, \hat{P}$

Define $u_t^{nj} \equiv \exp(V_t^{nj})$

$$\mu_{t+1}^{nj,ik} = \frac{\mu_t^{nj,ik} (\hat{u}_{t+2}^{ik})^{\beta/\nu}}{\sum_m \sum_h \mu_{t+1}^{nj,mh} (\hat{u}_{t+2}^{mh})^{\beta/\nu}} \quad (17)$$

$$L_{t+1}^{nj} = \sum_i \sum_k \mu_t^{ik,nj} L_t^{ik} \quad (18)$$

$$\hat{u}_{t+1}^{nj} = \frac{\hat{w}_{nj}}{\hat{p}_n} \left(\sum_i \sum_k \mu^{nj,ik} (\hat{u}_{t+2}^{ik})^{\beta/\nu} \right)^\nu \quad (19)$$

Dynamic Hat Algebra

Define $u_t^{nj} \equiv \exp(V_t^{nj})$

$$\mu_{t+1}^{nj,ik} = \frac{\mu_t^{nj,ik} (\hat{u}_{t+2}^{ik})^{\beta/\nu}}{\sum_m \sum_h \mu_{t+1}^{nj,mh} (\hat{u}_{t+2}^{mh})^{\beta/\nu}} \quad (20)$$

$$L_{t+1}^{nj} = \sum_i \sum_k \mu_t^{ik,nj} L_t^{ik} \quad (21)$$

$$\hat{u}_{t+1}^{nj} = \frac{\hat{w}_{nj}}{\hat{P}_n} \left(\sum_i \sum_k \mu_t^{nj,ik} (\hat{u}_{t+2}^{ik})^{\beta/\nu} \right)^\nu \quad (22)$$

Input $\beta, \nu, L, \hat{w}, \hat{P}$ Output $L', \mu, \hat{u}$

Algorithm

- 1 Estimate parameters of G to match static trade shares as in LR
- 2 Guess a path for $\{\hat{u}_t^{ik}\}$ that converges to 1
- 3 Use Equation 20 and initial migration flows to solve for the full migration path
- 4 Use Equation 21 to solve for the labor path
- 5 Compute the change in real wages using the static hat algebra equations
- 6 Use Equation 22 to compute the implied path for $\{\hat{u}_t^{ik}\}$
- 7 Iterate 2-6 until convergence

Conclusion

- Our framework synthesizes two frontier models:
 - Caliendo et al. (2019) for heterogenous workers
 - Lind and Ramondo (2023) for nested productivity
- As a result, the effect of tariffs is heterogeneous across workers **and** products
- Analyses:
 - To what extent can tariffs bring back domestic jobs?
 - To what extent to domestic workers in targeted industries benefit?
 - What is the cost to workers in other industries?