

Time to Build and Fluctuations in Bulk Shipping

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This Paper

How important are production lags in affecting industry dynamics?

- Broad literature on investment under uncertainty with adjustment costs.
- But little empirical evidence on quantitative importance of specific mechanisms.

Main result

- Adjustments cost in the form of time to build affect 1st and 2nd moments: reduces entry and fleet levels and volatility, and raises prices of trips.

Empirical strategy

- second-hand sale prices of ships can be used to non-parametrically estimate value functions and then recover per-period profits and entry/exit costs.
- “mirror-image” of Bajari, Benkard, Levin (2007).

Environment

A ship is a firm and its state is its age j and the industry state $x_t = (s_t, b_t, d_t)$

- $s_t = [s_t^0, \dots, s_t^A]$ is the industry age distribution, $b_t = [b_t^1, \dots, b_t^{\bar{T}}]$ is the backlog of ship orders, d_t is a demand shifter for trips (exogenous 1st order Markov).

Upon entry, ships are assigned a time to build $T_t(s_t, b_t, d_t)$. Timing is as follows:

1. Exit/entry decisions are made simultaneously.
 - Incumbents observe scrap value $\phi \sim F_\phi$ every period and decide to stay/exit.
 - Potential entrants observe entry cost $\kappa(s_t, b_t, d_t)$ and decide to stay out/enter (first empirical paper to allow for non-iid state dependent entry costs).
2. Incumbents receive per-period profits $\pi_j(s_t, d_t)$, even if they've decided to exit.
3. Entry/exit decisions are implemented and the state evolves.

- Value of an age j incumbent at t before observing ϕ

$$V_j(s_t, b_t, d_t) = \pi_j(s_t, d_t) + \beta E_\phi \max\{\phi, VC_j(s_t, b_t, d_t)\}$$

where $VC_j(s_t, b_t, d_t) \equiv E[V_{j+1}(s_{t+1}, b_{t+1}, d_{t+1})|s_t, b_t, d_t]$ and $E[\cdot]$ is an expectation over entrants N_t and age j exiters Z_t^j .

$$\begin{aligned} s_{t+1}^0 &= b_t^1 \\ s_{t+1}^j &= s_t^{j-1} - Z_t^j & j = 1, \dots, A \\ b_{t+1}^i &= b_t^{i+1} & i \neq T_t, \bar{T} \\ b_{t+1}^{T_t} &= b_t^{T_t+1} + N_t, \quad b_{t+1}^{\bar{T}} = 0 & T_t < \bar{T} \\ b_{t+1}^{\bar{T}} &= b_t^{\bar{T}} + N_t & T_t = \bar{T} \end{aligned}$$

- Value of entry is $VE(s_t, b_t, d_t) \equiv \beta^{T_t} E[V_0(s_{t+T_t}, b_{t+T_t}, d_{t+T_t})|s_t, b_t, d_t]$ and there is an entry cost $\kappa(s_t, b_t, d_t)$.

Equilibrium assumptions:

1. Potential entrants play a symmetric mixed entry strategy $\implies N_t$ is well approximated by a Poisson with entry rate $\lambda(s_t, b_t, d_t)$.
2. Equilibrium is symmetric \implies incumbents use a common cut-off rule for exit, which induces an exit probability $\zeta(s_t, b_t, d_t)$.

Equilibrium

Let $V_j(x; \zeta', \zeta, \lambda)$ be the incumbent's payoff from playing ζ' given others play ζ .

$$V_j(x; \zeta', \zeta, \lambda) = \pi_j(s_t, d_t) + \zeta' \beta E(\phi | \phi > VC_j(x; \zeta, \lambda)) + (1 - \zeta') \beta VC_j(x; \zeta, \lambda)$$

An equilibrium is an entry strategy λ and an exit strategy ζ such that

- Incumbent firm strategies represent a MPE.

$$\sup_{\zeta'_j(x) \in [0,1]} V_j(x; \zeta', \zeta, \lambda) = V_j(x; \zeta, \zeta, \lambda) \quad \forall j \in \{0, 1, \dots, A\}, x \in (S \times B \times D)$$

- Entrants have zero expected profits or entry rate is zero (or both).

$$VE(x) \leq \kappa(x) \quad \forall x \in (S \times B \times D) \quad \text{with equality if } \lambda(x) > 0$$

1. Second-hand sales data → used to estimate value functions.
2. Contracts data on prices and quantities of trips → used to estimate trips demand.
3. Ship orders, deliveries, demolitions → used to estimate state transitions.

Main challenge in dynamic games is computing continuation values.

- Second-hand sale prices are used to recover value functions nonparametrically
⇒ need to assume homogenous agents, no private info, no search frictions, so that sale price = ship value.

Two step procedure:

1. Estimate value functions, state transitions, and exogenous objects not determined in equilibrium (d_t and T_t) “outside” of the model.
2. Given estimates in 1, use the model’s restrictions to estimate primitives: profits, scrap value distribution, entry costs.

Step 1

- Estimate demand for trips using supply shifters (fleet size/age, fuel costs).
- Estimate $V(\cdot)$ using local linear regression on re-sale prices and grid $[j, X]$.

$$\min_{\beta_0, \beta_j, \beta_S, \beta_B, \beta_d} \sum_i \{ p_i^{SH} - \beta_0(j, X) - \beta_j(j, X)(j_i - j) - \beta_S(j, X)(S_i - S) \\ - \beta_B(j, X)(B_i - B) - \beta_d(j, X)(d_i - d) \}^2 K_h([j_i, X_i] - [j, X])$$

- Estimate TTB running OLS of T_t on X_t .
- Estimate transition of X_t assuming conditional independence between N_t, Z_t, ϵ_t

$$Pr(x_{t+1}|x_t) = Q(x_t, x_{t+1}) f_N(N_t = n|x_t) f_Z(Z_t = z|x_t) f_\epsilon(\epsilon_{t+1})$$

Project entry/exit rates ζ_t, λ_t on X_t to compute conditional densities $f_N(\cdot|X_t), f_Z(\cdot|X_t)$. Use demand parameters to compute ϵ_{t+1} .

Step 2

- Given P, V from step 1 we can compute $VC_A = PV_A$, and if we assume only age A ships may exit then the share of exiters is $\frac{Z_t}{s_t^A} = 1 - F_\phi(VC_A(X_t))$. We then run local linear regression at all points of interest ν .

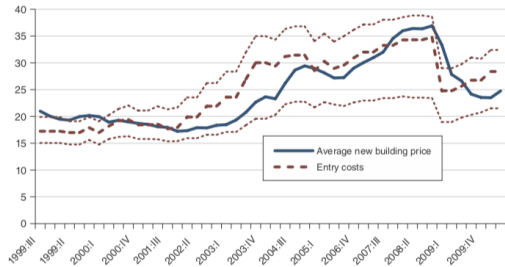
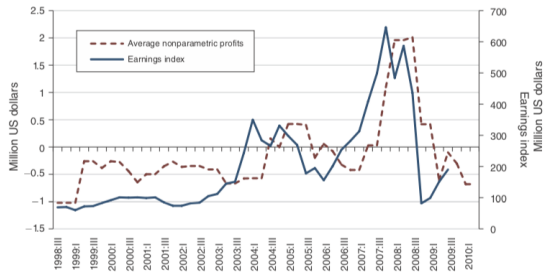
$$\min_{\beta_0, \beta_{VC}} \sum_t \left[\frac{Z_t}{s_t^A} - \beta_0(\nu) - \beta_{VC}(VC_A(X) - \nu) \right]^2 K_h(VC_A(X) - \nu)$$

so that $1 - F_\phi(\nu) = \beta_0(\nu)$.

- Profits backed out as $\pi_j = V_j - \beta VC_j$ and entry costs backed out from free entry condition

$$\beta^{T(X)} P_X^{T(X)} V_0 = \kappa(X)$$

Model fit



Counterfactuals

- Profits are parametrized by cost parameters θ .
- Under perfect competition in the spot market for trips, firms optimize and obtain $\pi(\theta) \implies \theta$ can be estimated via NLLS using non-parametric $\hat{\pi}$ from step 2.
- We can now compute counterfactual equilibria, i.e., solve for λ, μ, V given primitives θ, κ, F_ϕ . Two counterfactuals considered: constant and no TTB.
- Experiment 1: impulse response to a positive demand shock.
- Experiment 2: long-run industry dynamics.

Impulse response to positive demand

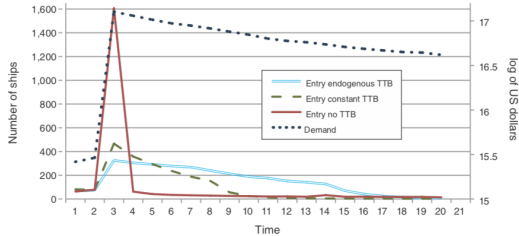


FIGURE 14. ENTRY UNDER ENDOGENOUS, CONSTANT, AND NO TIME TO BUILD (TTB)

Note: Demand depicted on right axis.

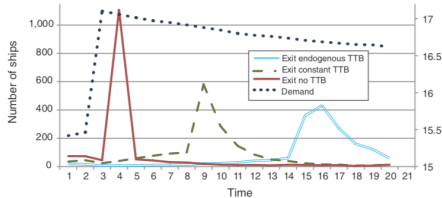
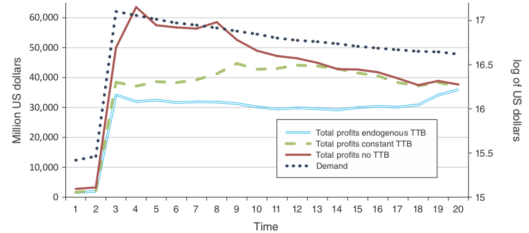


FIGURE 15. EXIT UNDER ENDOGENOUS, CONSTANT, AND NO TIME TO BUILD (TTB)



Panel B

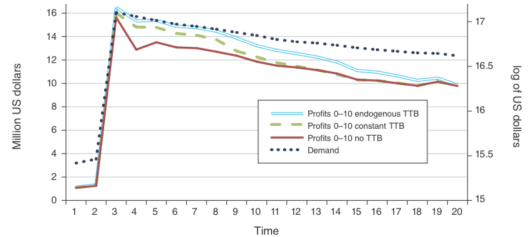


FIGURE 18. TOTAL (sum of) AND PER SHIP PROFITS FOR ALL AGE GROUPS UNDER ENDOGENOUS, CONSTANT, AND NO TIME TO BUILD (TTB)

Long run dynamics

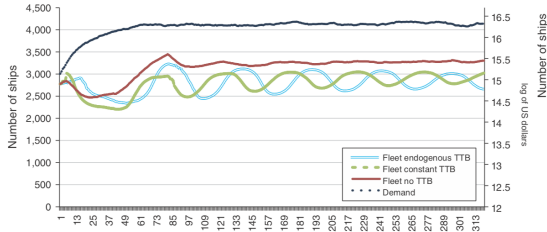


FIGURE 20. AVERAGE FLEET FROM LONG-RUN SIMULATIONS

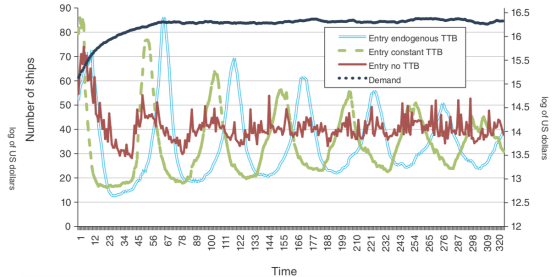


FIGURE 21. AVERAGE ENTRY FROM LONG-RUN SIMULATIONS

- Echo effects vanish quickly when there is no TTB

Methodology

- with homogenous agents, observed transaction prices of a good whose value is the solution to a dynamic problem can be used to estimate dynamic parameters.

Economics

- Much work on investment under uncertainty considers comparative statics in demand uncertainty, broadly defined.
- But then what specific policies should we think of to affect demand uncertainty, especially in the context of trade?
- This paper quantifies a specific channel which affects investment, so we can think of obvious policy implications (e.g. building more shipbuilding docks).