Discussion of Industrial policies for multi-stage production: The battle for battery-powered vehicles

(Head, Mayer, Melitz, Yang)

Trade, value chains and financial linkages in the global economy BDI / ECB / WB

November 2024

Rocco Macchiavello LSE • This paper develops a quantitative method to analyse industrial policies for multi-stage production and applies it to the EV industry

Contribution I

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Contribution II

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My 2 cents on

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Methodological Contribution

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- Help a(n uninformed) reader in appreciating the contribution Section 4:
 - Discussion of literature in intro refers to *interdependencies*, *sub* and *super* modularity. Is it worth formally defining these concepts in the context of your model?
 - Does your model nest models in the literature? You could be explicit about which dimensions it generalizes existing models. (Maybe even deploy your algorithm to solve those simpler models)

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Section 5:

- Currently it shows that the approach is feasible in a demanding environment ("it can be done, but results are complex")
- This invites the reader to be skeptical about policy counterfactuals later on.
- Would exploring a continuum of simulations illustrate how [some] key parameters matter?
 Maybe you could do that with the simpler model you eventually take to the data.
- This would help the reader build [some] intuition and might be cheaper than tackling the robustness of counterfactuals to estimates.

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- The data are impressive and novel.
 - Under mild assumptions, it allows us to track the supply chain (location, factory) for many EVs across multiple stages and countries. I *really* like this aspect.
 - [More generally: enormous gains from global industry studies in btw IO & Trade]
- It seems an important computational constraint is the sheer number of potential *paths* a firm could choose from to optimize
 - The authors make very sensible choices once they get to estimation
- There is not a lot of discussion of whether the potential sets of paths available to a firm could be more limited in practice
 - MNCs entry & location decisions process
 - Path-dependency
 - ... ?
- (Other data might also be useful for validation/estimation)

Estimation

- There is <u>a lot</u> to absorb here. What I understood:
 - a. Taking entry in market *n* as given, we focus on the choice of path i.e., sourcing choice for the two stages (assembly, battery)
 - b. The probability of a given path can be expressed as the product of the conditional probabilities at each stage.
 - c. In turn, each of those, is a function of variable and fixed costs at that stage.
 - d. In a first step, we can estimate the probability of each stage separately. This yields estimates for the variable costs.
 - e. In the second step, SMM can be used to recover also the fixed costs.
- Assumptions required to go c. → d. not entirely clear to me. It would be helpful perhaps to spend a bit more time there.
 - Is the idea that the bilateral fixed-effects for origin and destination at the relevant stage absorb the fixed costs (?)
- I *really* liked benchmarking of cells' cost share to industry "*estimates*" (even s.t. caveat). Can "insider" knowledge be leveraged more systematically?
 - E.g., as constraints/targets to estimate the model, or for validation.

Taking the Model to the Industry

- The tool developed in the paper can be applied to (*m*)*any* industry.
- For example, a great fit for the apparel industry:
 - Brands selling in multiple countries, deciding where to do design, and de-localize production,
 - Many buyers, infinitely many suppliers,
 - Industry in stead-state, little technological change & dynamic learning effects
- But is this a good model for the EV industry?
 - Some stylized facts consistent with the model. However ...

No.	Manufacturer	# Markets	# Models	Sales	Sales-exCHN
				Cum. Share (%)	Cum. Shr (%)
1	Tesla	23	4	20.5	28.0
2	Volkswagen	24	23	30.0	42.4
3	Hyundai	23	15	35.6	54.1
4	Stellantis	18	19	40.3	63.7
5	BMW	24	7	44.0	69.7
6	Renault	19	6	46.7	75.3
7	Mercedes-Benz	24	9	49.3	80.0
8	Geely	24	15	55.6	84.4
9	Ford	22	4	57.5	88.3
10	Nissan-Mitsubishi	23	8	60.1	91.9
11	SAIC	17	21	72.8	95.0
12	General Motors	7	7	74.1	96.6
13	Toyota	23	8	74.9	97.8
14	Rivian	3	3	75.3	98.6
15	BYD	10	14	89.6	98.9

Table 5: 15 top firms in 2022

Note: Market shares defined over 24 countries used in simulations.

- In-sample global *HHI* ≈1375 (*C4* 63.7)
- In US, *HHI* (used to be) >2500, in EU *HHI* >1000 (*C4* ≈54)
- Concentrated market, with substantial dynamic
- Batteries (and patents on batteries) even more concentrated

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 - EV industry is (quite) highly concentrated at both stages (autos, batteries) \rightarrow bilateral oligopoly
 - Industry clearly not in steady state: motivating facts confirm this.
 - Massive technological progress on batteries
 - 1. Cost of batteries is rapidly declining
 - 2. Battery quality (milage, charging times, safety) will determine who wins the race!
- Maybe discuss a bit more the trade-offs involved in modelling choices + possibly implications for the evaluation of policy counterfactuals.

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• We need more work like this at the intersection of trade & IO

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Thank you!