

# Product Innovation, Product Diversification, and Firm Growth: Evidence from Japan's Early Industrialization

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# Firm Growth and Product Line Expansion

Lots of evidence that firms usually grow by adding product varieties

- Most empirical work looks at manufacturers
- But some for services too (e.g., Hsieh and Rossi-Hansberg's "Industrial Revolution in Services" —geographic expansion is one type of variety)
- Famous anecdotes: Google Alphabetizes, Amazon expands

# Firm Growth and Product Line Expansion

All frameworks assume a certain degree of symmetry among products

- In expectation, all potential product introductions are equally effective channels for growth

But supply- and demand-side effects may be different

- Consumers' willingness to substitute and firms' economies of scope are driven by very different primitives

Innovative products may create spillovers that me-too products do not

Bottom line: It might not just be the number, but type, of products that matter for growth

# Firm Growth and Product Line Expansion

We have the data to treat products as different, and we find those distinctions matter

Our setting: Japanese cotton spinning industry 1893-1914

We find:

- Vertical and horizontal product expansions are different
  - Vertical: climbing up the technological ladder
  - Horizontal: making new varieties within firm's existing technology
- Vertical expansions are necessary ingredient in long-run firm growth
  - Vertical expansions necessary for *horizontal* product growth
- Horizontal expansions don't make vertical expansions easier
- Attempts at vertical expansions often fail and are retried
- Mechanisms related to greater flexibility and demand-side "appeal"

# Similar Cases/External Validity

- Our findings may not apply universally, but are empirically observed in a range of other contexts
  - Industry level
    - Global mobile phone industry: leaders jumped into high-end phones first, then captured the low-end market (even before smartphones!)
    - Robotics: strong impact of interaction between what they call “new knowledge” and “adding depth to existing knowledge” on the diversity of the product portfolio
  - Firm level
    - Honda: used race cars as a springboard for consumer car market
    - Shimano: brand appeal of high-end drivetrains translated into success in the low-end market
    - TSMC: production flexibility by reusing obsolete high-end capacity

# Conceptual Framework

(Not trying to break new theoretical ground; borrows heavily from McCardle (1985) and Jovanovic (1982); just making sense of our findings)

- Firms endowed with growth potential but know only the prior distribution
- Can run costly product upgrade trials to learn more
- Entering trials involves a fixed cost (new machines, engineers, etc.)
  - Only select firms enter trials (selection treated separately)
- Conditional on entering trials, these succeed or fail
  - Successful trials boost growth and knowledge/brand appeal
  - Knowledge/brand appeal useful across the spectrum
- Too many unsuccessful trials  $\Rightarrow$  exit trials, join firms that never tried
- Enough successful trials  $\Rightarrow$  exit trials and grow through portfolio diversification (apply accumulated knowledge/brand appeal to low-end products)

# Conceptual Framework: Empirical Implications

- *Sorting pattern—three firm types in the long run:*
  1. No product upgrading (empirically, about half of the sample)
  2. Mostly failed product upgrade trials (introduced high-end machines but failed to diversify, eventually acquired by type-3 firms)
  3. More successful product upgrade trials, moved to product diversification (became fast-growing firms, serial acquirers)
- *Product varieties and firm growth*
  - Past upgrade trials predict growth only marginally (mix of type-2 and type-3 firms)
  - Subsequent product diversification isolates high-growth firms  $\Rightarrow$  strong positive growth effect of the interaction term
- *Output cuts as a source of exogenous variation*
  - Coincidence of industry-wide mandatory output cuts on low-end products and the arrival of high-end machine orders reduced opportunity costs of upgrade trials
    - Neither can be timed, so pretty much an exogenous cost-shifter

# Setting and Data

Japan's cotton spinning industry was the first modern manufacturing industry in Asia; at world frontier by 1915-20

Supply chain:

Raw Cotton → Cotton Spinning → Textile Weaving → Textile products

Output is “yarn” (read: thread)

Two important attributes: count (fineness) and finish

- Count: units of yarn length that have a given weight
  - Higher counts mean finer (higher quality) yarn
- Finish: the way threads are spun
  - Right-twist and left-twist are standard and treatable as equivalent
  - Doubled and gassed yarns are higher quality



# Setting and Data

## Monthly plant-level data

- Production in physical units of yarn by type (count-by-finish)
  - We consolidate 201 types in raw data to 35 types
  - We sometimes dichotomize yarn into “low-end” and “high-end”; high-end is above 20-count and/or doubled or gassed
- Input use (operating spindles, workers, raw cotton) in physical units
- Output prices (for select counts)
- Wages
- Existence of industry-mandated output restrictions

## Semiannual plant-level data

- Machine installed capacity, orders, and installations
- Number of engineers
- Composition of firm’s board of directors







# Setting and Data

## Key definitions

- Product upgrade/vertical expansion is making new yarn type that is:
  - High-end AND
  - Has higher count than any the firm previously made “at scale”
    - At scale: accounting for at least 3% of firm’s output
- Product diversification: making a new yarn at scale at a count lower than or equal to any count the firm previously made
- Product trials: making a yarn in period  $t$  that a) the firm did not make in  $t - 1$  (semiannual periods), b) is not made at scale in  $t$ , and c) had never been made at scale by the firm before  $t$ 
  - Trials can succeed—grow to reach production at scale
  - Trials can fail but be retried later

# Trials

## Firm-by-product-line production episodes

	All (1)	Fraction of total	Never scaled (2)	Ratio (2)/(1)
New product lines	685	1.000	271	0.396
Of which: never a trial initially trial products	246 439	0.359 0.641	271	0.617
Of which: Upgrade lines	76	0.111	33	0.434

## Experiments

	All	Successful (scaled)	Failed (not scaled)	Fraction failed
All trial products	819	223	596	0.728
Of which: upgrades diversifications	116 703	42 181	74 522	0.638 0.743
fraction upgrades	0.142	0.188	0.124	

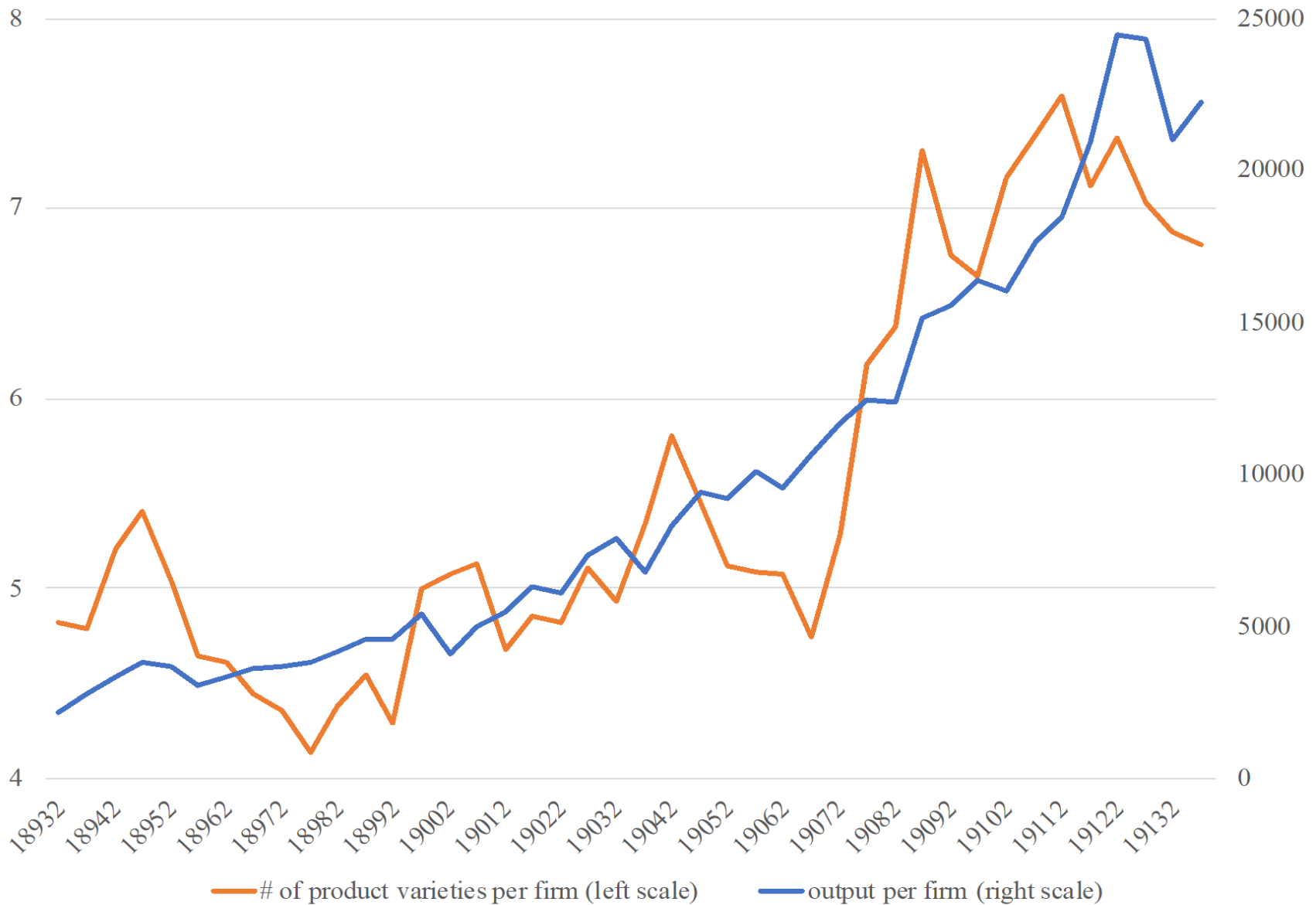
# Trials

Trials usually fail, two-thirds to three-fourths of the time

Given that trials are small by definition and often fail, they themselves cannot be a source of growth

However, we will show that trials are related, probably causally, to firm growth through product expansion

# Firm Growth and Product Expansion



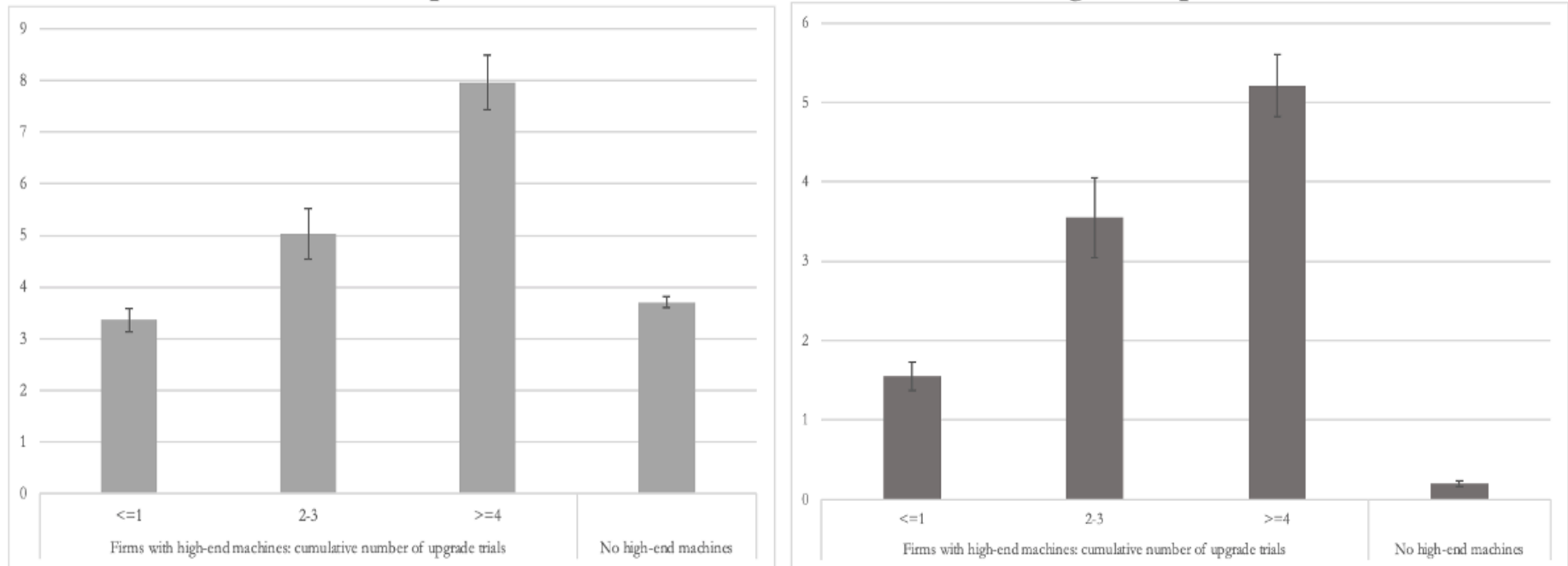
# Upgrade Trials and Product Expansion

Figure 2.

Panel A. Cumulative upgrade trials and number of low-end and high-end products in firms' portfolios

Low-end products

High-end products



From regression analysis: Cumulative upgrade trials 25<sup>th</sup> %ile = 1, 75<sup>th</sup> %ile = 5

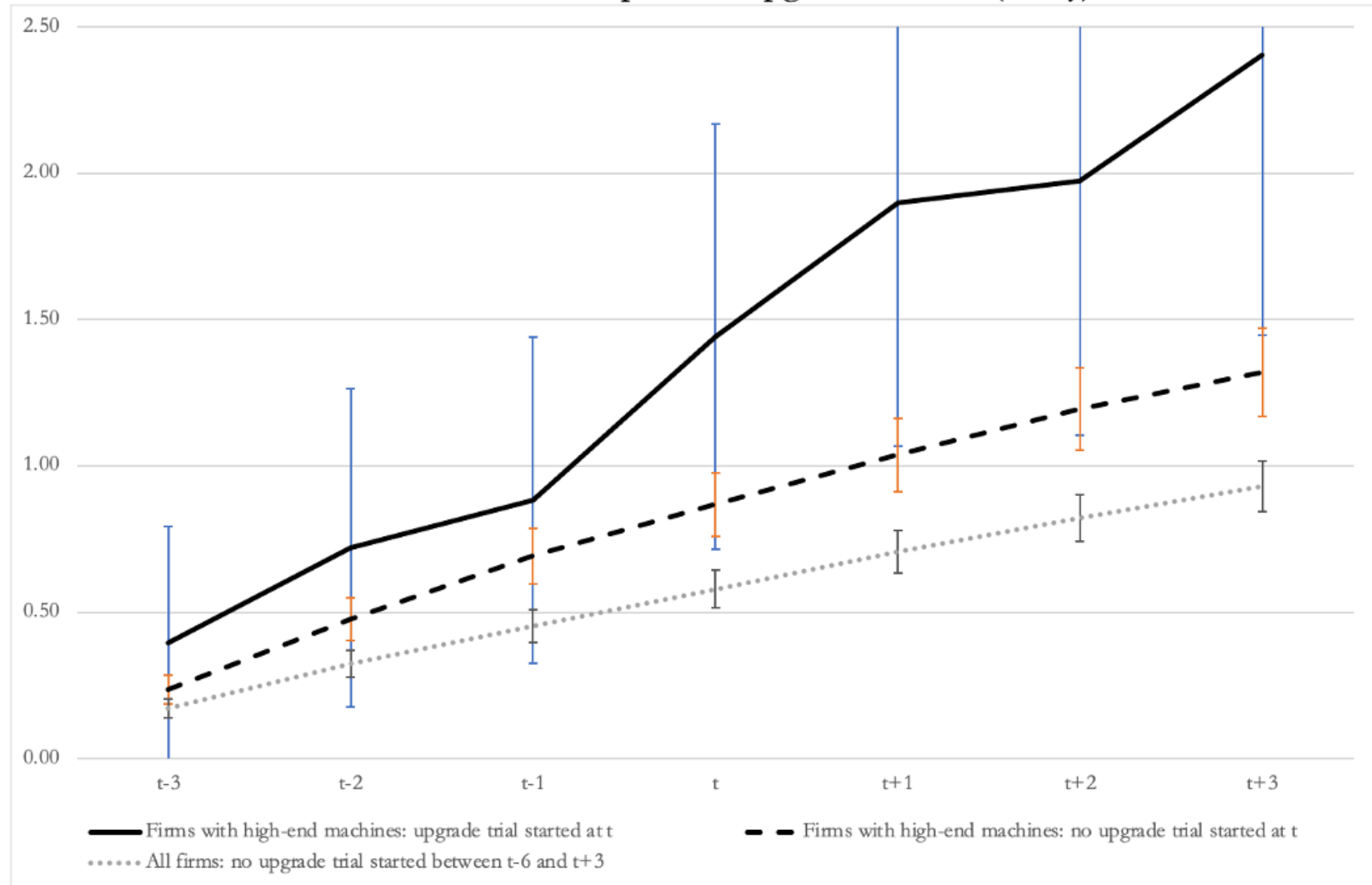
IQR tied to about 1.7 (0.43x4) additional new varieties



# Upgrade Trials and Product Expansion (cont.)

Figure 2.

Panel B. Dynamics of cumulative new-to-firm horizontally diversified products from  $t-3$  to  $t+3$  after a product upgrade trial at  $t$  (if any)



# What Drives Product Upgrade Trials?

Number of UPGRADE experiments in  $t$ :

		All Firms		Firms with high-end machines	
I[had high-end machine in $t$ ]	1.066*** (0.378)				
I[installed high-end machinery in $t$ ]		1.387*** (0.305)	0.700** (0.379)	0.943*** (0.343)	0.295 (0.362)
I[installed high-end machinery in $t$ ] x I[mandated output cuts in effect in $t$ ]			1.111* (0.588)		1.846*** (0.603)
I[installed low-end machinery in $t$ ]		-0.002 (0.334)	-0.132 (0.374)	-0.039 (0.420)	-0.045 (0.431)
I[employs univ.-educated engineer in $t$ ]			0.565 (0.421)		0.485 (0.494)
I[has exchange merchant on board in $t$ ]			1.455*** (0.416)		1.666*** (0.547)
Controls: period FEs, “age” dummies	Yes	Yes	Yes	Yes	Yes
N	1,618	1,618	1,618	701	701

# What Drives Product Upgrade Trials?

Number of DIVERSIFICATION experiments in  $t$ :

		All Firms		Firms with high-end machines	
I[had high-end machine in $t$ ]	0.941*** (0.209)				
I[installed high-end machinery in $t$ ]		1.099*** (0.210)	0.830*** (0.300)	0.746*** (0.212)	0.582* (0.297)
I[installed high-end machinery in $t$ ] x I[mandated output cuts in effect in $t$ ]			0.059 (0.423)		0.288 (0.424)
I[installed low-end machinery in $t$ ]		0.236 (0.215)	0.148 (0.218)	0.138 (0.260)	0.072 (0.257)
I[employs univ.-educated engineer in $t$ ]			0.743*** (0.177)		0.675*** (0.211)
I[has exchange merchant on board in $t$ ]			0.387** (0.163)		0.442* (0.268)
Controls: period FEs, “age” dummies	Yes	Yes	Yes	Yes	Yes
N	1,618	1,618	1,618	701	701

# High-End Machines

	Firms with high-end machines	Firms without high-end machines
Fraction of firms with merchant board member	0.71	0.57
Number of university-educated engineers	1.61	0.14
Number of technical-college-educated engineers	4.76	0.72
Experimental product introductions per period: All	0.60	0.25
Of which, fraction successful	22%	34%
Experimental product introductions: Upgrade	0.10	0.03
Of which, fraction successful	32%	36%
Experimental product introductions: Diversification	0.50	0.22
Of which, fraction successful	18%	35%
Non-experimental product introductions: All	0.11	0.11
Non-experimental product introductions: Upgrade	0.06	0.01
Non-experimental product introductions: Diversification	0.05	0.09

# What Drives Product Upgrade Trials? An IV

An IV for upgrade trials: interact a) installing high-end machinery in  $t$  and b) imposition and magnitude of mandated output cuts in  $t$

- Relevance condition
  - High-end machinery correlated with making high-end products; output cuts imposed on low-end product types free up firm resources to try new things
- Exclusion condition
  - Delivery time lags of high-end machine orders are 1-2 years (sometimes longer) and uncertain
    - Unlikely firm can anticipate imposition or size of mandated output cuts that far in the future
  - Mandated cuts from aggregate demand fluctuations unlikely to be tied to shifts in firms' unobservable innovative capabilities

# Product Upgrade Trials and Growth

	DV: number of upgrade experiments started at $t$		DV: $\Delta \ln(\text{output})$	
	First stage	Placebo test	Second stage	
Cumulative upgrade experiments			0.012 (0.012)	-0.014 (0.014)
Fraction of products that are low-end	-2.371*** (0.463)	-2.360*** (0.461)	-0.029 (0.027)	-0.080** (0.039)
(Cumulative upgrade experiments) x fraction low-end products				0.081** (0.034)
(Fraction output subject to cuts) x $\Delta \ln(\text{installed high-end spindles})$	2.539*** (0.647)			
(Fraction output subject to cuts) x $\Delta \ln(\text{installed low-end spindles})$		-0.040 (0.565)		
Controls: $\Delta \ln(\text{installed spindles})$ , engineer and merchant board member dummies, logged output, period FEs, “age” dummies	Yes	Yes	Yes	Yes
N	1,608	1,608	1,608	1,608
Estimation	Poisson	Poisson	IV	IV

Additional upgrade experiment for firm at mean fraction of low-end products raises growth rate by about 5.5% (one-third of IQR)

# Growth Mechanisms: Production Flexibility

Measure in monthly data how often a firm rebalances its product portfolio

Changes in “lead direction” of twists within counts

- E.g., firm had been producing 80% left-twist and 20% right-twist 16-count and then in next month shifts to 30% left-twist and 70% right-twist

Changes in “lead count” within finishes

- E.g., firm had been producing 80% 16-count and 20% 20-count and then in next month shifts to 30% 16-count and 70% 20-count

We focus on these changes for low-end products

# Growth Mechanisms: Production Flexibility

	Number of changes in lead direction	Number of changes in lead count	$\Delta \ln(\text{output})$	$\Delta \ln(\text{output})$
Cumulative upgrade experiments	0.241*** (0.083)	0.153* (0.085)		
Cumulative diversification experiments	0.022 (0.016)	0.012 (0.024)		
Number of changes in lead direction			0.010** (0.004)	
Number of changes in lead count				0.012*** (0.004)
I[employs univ.-educated engineer in $t$ ]	0.190 (0.152)	-0.091 (0.293)	0.103** (0.041)	0.103** (0.041)
I[has exchange merchant on board in $t$ ]	0.084 (0.098)	0.154 (0.170)	0.011 (0.026)	0.008 (0.025)
I[installed high-end machinery in $t$ ]	0.184 (0.141)	0.258** (0.122)	0.014* (0.007)	0.014* (0.007)
I[installed low-end machinery in $t$ ]	0.171 (0.127)	-0.003 (0.111)	-0.002 (0.015)	-0.002 (0.015)
Total output	-0.033 (0.035)	-0.138*** (0.038)	-0.298*** (0.048)	-0.296*** (0.047)
N	1,605	1,605	1,608	1,608
Period and firm FEs	Yes	Yes	Yes	Yes

Add'l upgrade experiment tied to increase in lead direction (count) of half (one-eighth) its mean



# Growth Mechanisms: Demand Appeal

Measure quality of *low-end* products

Quality metric: Khandalwal (2010)-style demand estimation

- Essentially, product's market share after controlling for price differences

We don't have price data for every product, but we do for a key 20-count yarn that accounted for around  $\frac{1}{4}$  of industry output

IV for price: lead-count changes production flexibility measure from above, interacted with output controls (imposed on yarns up to 20-count)

- Intuition: flexible firms could more easily substitute to counts above 20 when output controls were in place, and as such did not need to reduce 20-count prices as much
- But not lead-twist—shouldn't have effect on pricing

# Growth Mechanisms: Demand Appeal

## Demand Estimation

	DV: ln(20-count price)		DV: ln(share)
	First stage	Placebo test	Second stage
ln(20-count price)			-5.407 (6.136)
Lead-count changes	-0.002** (0.001)		
(Lead-count changes) x (mandated output cuts)	0.032*** (0.007)		
Lead-twist changes		-0.001 (0.001)	
(Lead-twist changes) x (mandated output cuts)		0.005 (0.006)	
Controls: Period and firm FEs	Yes	Yes	Yes
N	743	743	743

# Growth Mechanisms: Demand Appeal

	DV: Quality			
	OLS	OLS	IV	IV
Cumulative upgrade experiments	0.067*** (0.026)	-0.037 (0.081)	0.503*** (0.062)	-0.101 (0.217)
(Cumulative upgrade experiments) x (fraction of low-end products)		0.158 (0.114)		0.925*** (0.320)
Fraction of low-end products	0.884*** (0.289)	0.803*** (0.294)	1.532*** (0.289)	1.028*** (0.364)
I[employs univ.-educated engineer in $t$ ]	0.989*** (0.118)	1.001*** (0.119)	0.763*** (0.116)	0.762*** (0.117)
I[has exchange merchant on board in $t$ ]	0.189* (0.112)	0.167 (0.112)	0.027 (0.107)	-0.056 (0.110)
Controls: $\Delta \ln(\text{installed spindles})$ , period FEs, “age” dummies	Yes	Yes	Yes	Yes
N	721	721	721	721

# Robustness: Mean Growth or Variance of Growth?

Firms that make high-end products are more likely to survive to the end of the sample

	Survivors	Exit (acquired)	Exit (liquidated)	Total
Had high-end machines	19	22	1	42
No high-end machines	14	31	18	63
Total	33	53	19	105

# Conclusions and Discussion

Product expansion and firm growth go together in our sample, as the literature has found in other settings

However, all product-line expansions are not the same

Clear asymmetry in product expansion and growth patterns in our sample

High-growth firms went outside their existing technological frontiers and experimentally introduced innovative products

- This led to growth not just in high-end products, but low-end as well
- Low-end/diversification experiments are not related to long-run growth
- Mechanism behind high-end-driven growth seemingly related to knowledge gains in manufacturing flexibility and improvements in some notion of demand-side appeal