

Allocating Talent in Organizations: Evidence from Quasi-Random Exits of Physicians

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This paper

- How well can organizations allocate tasks across heterogeneous workers when individual quality is not perfectly observed? ([Holmstrom and Milgrom, 1991](#); [Holmström, 2017](#))
- One way to approach this is to investigate whether, after the loss of talent, organizations reallocate work in ways that preserve quality composition, or shocks pass through mechanically
- We develop an empirical framework that uses **quasi-random personnel exits** and **observational measures of quality** to study such [ability-targeting inside organizations](#)
- Applies the framework to data on Brazilian hospitals, where patient assignments and physician outcomes are observed at scale
 - ▶ Computes risk-adjusted observational survival rates as a measure of physician quality
 - ▶ Uses unexpected deaths of physicians as exogenous shocks to the available set of clinical talent

Preview of framework and core result

- Think about the quality composition of an organization as the **weighted average of its workers' *intrinsic* quality**
- **Once someone leaves**, what happens to this measure depends on how good that worker was, their task share, and who takes over their tasks
- Our framework contrasts **realized change** in **quality composition** to the **mechanical change** created by the exit *absent any quality-targeted reallocation*
 - ▶ Simple test of how “blind” organizations are to differences in worker quality
- This relationship can be summarized by a single parameter, estimated in a DiD relating exit-induced quality shocks to changes in the quality composition of workers
 - ▶ $\beta = -1$ under one-to-one pass-through and $\beta = 0$ under full quality-preserving reallocation

⇒ We find $\hat{\beta} = -0.38$

Outline

Setting and Data

Framework and Empirical Strategy

Main Results

Signal Reliability and Impacts on Survival

Conclusion

Brazilian Universal Healthcare System – SUS

- Brazil's healthcare system (SUS) ensures **universal access to healthcare**
 - ▶ Largest publicly funded healthcare system in the world
- SUS provides both preventive and curative care free of charge through public institutions and a network of contracted private facilities

Data

- **Universe of inpatient admissions** from 2008 to 2024 (*SIH-Datasus*)
 - ▶ Observe clinical and sociodemographic characteristics of patients, procedures, discharge, and mortality
 - ▶ We also observe the [identity of the physician responsible for the case](#)
- **Physician-level administrative data** (*Federal Council of Medicine/CNES-Datasus*)
 - ▶ Demographics, professional background, workplaces, and exact date of death

Sample

- We have data on 315,902 physicians, of whom 8,198 die during the period
 - ▶ Many deaths did not come as a surprise
 - ▶ We focus on **physicians *actively working* in the month of death**: 1,021 deaths affecting 1,200 hospitals (out of 7,700)

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- Final sample uses **affected and unaffected hospitals *matched* (1:1)** on pre-event size and characteristics
 - ▶ Because deaths are rare and *mechanically* more likely in hospitals with larger physician workforces (≈ 100)
- **Analysis is at the *hospital-area-of-care level*** as hospitals operate through semi-autonomous clinical service lines
 - ▶ Areas of care are defined by ICD-10 subchapters, such as ischemic heart diseases

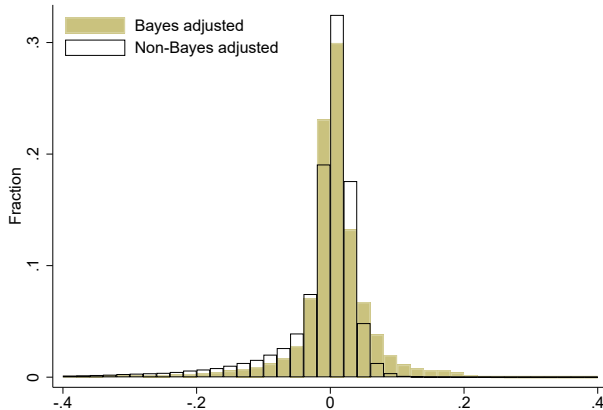
Measuring physician quality

- We estimate **risk-adjusted physician survival value-added** within areas of care s :

$$\text{Survival}_{it} = \theta_{js} + X'_{it}\Theta + \varepsilon_{it}.$$

- Controls include patient demographics, clinical characteristics, recent admissions, comorbidities, hospital fixed effects, and quarter-year fixed effects
- We refer to θ_{js} as *observational value-added*
 - ▶ Our approach doesn't require observational VA to equal true clinical quality, but it must provide an informative signal of it (we provide evidence that this is the case)
- Use **empirical Bayes** to account for differences due to sampling error

Variation in observational value-added



- SD $\approx$ 7.7 percentage points
- Balanced across alive and deceased physicians

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Framework: average clinical effectiveness

The starting point is the **quality composition of physicians** in a hospital–area, measured as their *patient-share-weighted average quality*:

$$\bar{\mu}_t = \sum_{j=0}^N \alpha_{jt} \mu_j.$$

- μ_j is physician j 's underlying *value-added for patient survival*
- α_{jt} is the share of cases assigned to physician j in period t
- We ask what happens to $\bar{\mu}_t$ when physician $j = 0$ exits

Framework: decomposing the effect of an exit

One can show the **change in quality composition** $\Delta \bar{\mu}$ following $j = 0$ exit can be decomposed into:

$$\Delta \bar{\mu} = \underbrace{\sum_{j=1}^N \left(\alpha_{j\text{post}} - \frac{\alpha_{j\text{pre}}}{1 - \alpha_{0\text{pre}}} \right) \mu_j}_{\text{internal reallocation}} + \underbrace{\alpha_k \mu_k}_{\text{external replacement}} - \underbrace{\alpha_{0\text{pre}} (\mu_0 - \bar{\mu}_{\text{pre}}^{-0})}_{\text{quality shock}} \quad (1)$$

1. **Internal reallocation:** how the caseload left is redistributed among incumbent physicians relative to a random reallocation benchmark
2. **External:** changes due to a new physician, with its magnitude reflecting the replacement's quality and the share of the caseload she absorbs
3. **Quality shock:** the **mechanical effect** of the exit **before any quality-targeted response**, given by the deceased's quality relative to peers $\times$ his baseline caseload share

Framework: mechanical shock versus organizational response

- After some algebra, the change in quality composition of physicians can be written as:

$$\Delta \bar{\mu} = (\gamma - 1) \times \alpha_{0,pre} \left(\mu_0 - \bar{\mu}_{pre}^{-0} \right).$$

- γ captures the hospital's overall endogenous response
- Benchmarks:
 - ▶ $\beta = \gamma - 1 = -1$: one-to-one mechanical pass-through
 - ▶ $\beta = \gamma - 1 = 0$: full mitigation of the quality shock

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⇒ For a given hospital-area, the **change in quality composition of physicians** before and after a physician's death is proportional to the size of the quality shock, scaled by $\beta = \gamma - 1$

From framework to data

- In the data, we observe a panel of hospital–area cells, together with the timing and intensity of the turnover-induced shocks. DiD framework:

$$\bar{\mu}_{hst} = \beta \times \alpha_{0hs,pre} \left(\mu_{0a} - \bar{\mu}_{hs,pre}^{-0} \right) \cdot \mathbb{1}\{t \geq \tau\} + \mathbf{FE}_{hs} + \mathbf{FE}_{ts} + \mathbf{X}'_{hat} \Omega + \varepsilon_{hst}. \quad (2)$$

- ▶ $\bar{\mu}_{hst}$ is the **quality composition of physicians** in hospital-area hs in quarter t
- ▶ Treatment is the interaction between the **quality shock** and the **post-death indicator**
- ▶ Unit fixed-effects $\mathbf{FE}_{hs}$
- ▶ Time fixed-effects $\mathbf{FE}_{ts}$ (area-specific so variation is within area across hospitals)
- ▶ $\mathbf{X}$ controls for interactions between time FE and $\alpha_{0hs,pre}$ to hold fixed the size of the vacancy so identification comes only from variation in the quality content of the vacancy

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 - ▶ $\mathbf{X}$ controls for interactions between time FE and $\alpha_{0hs,pre}$ to hold fixed the size of the vacancy so identification comes only from variation in the quality content of the vacancy
- As μ is unobserved, we use θ to obtain an estimable expression of equation (2) where $\mu_{hst} = \lambda\theta_{hst} + \eta_{hst} \rightarrow$ error in the estimating eq. will be a function of both ε and η

Identification

- Standard parallel-trends condition: **common trends in residual quality composition**
 - ▶ Unobserved determinants of average clinical value-added would have evolved similarly across areas differentially exposed to the quality shock
- Additional condition: **common trends in residual forecast errors**
 - ▶ More- and less-exposed areas should not experience differential shocks to the unobserved component of true quality that is invisible in observational value-added
 - ▶ We test for this by estimating effects on proxies for the forecast error (avg physician characteristics net of empirical quality) and find null effects ([Abaluck et al., 2021](#))
- Check for pre-trends in an analogous event-study specification of (2) with leads and lags

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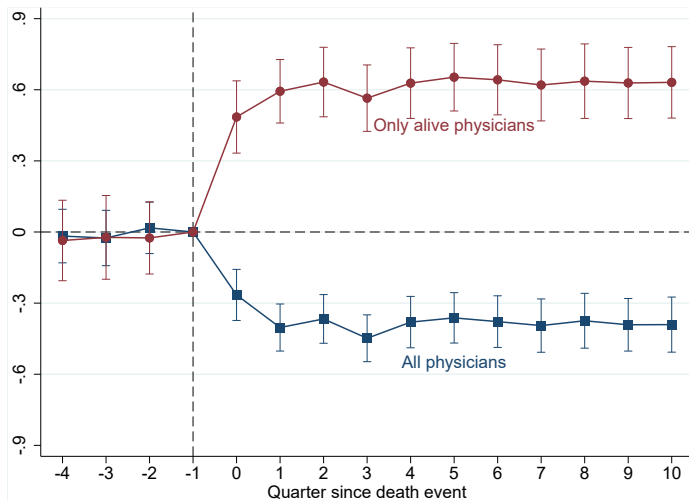
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Main result: quality shocks do not pass through one-to-one

Figure: Effects of Quality Shocks on Quality Composition of Physicians

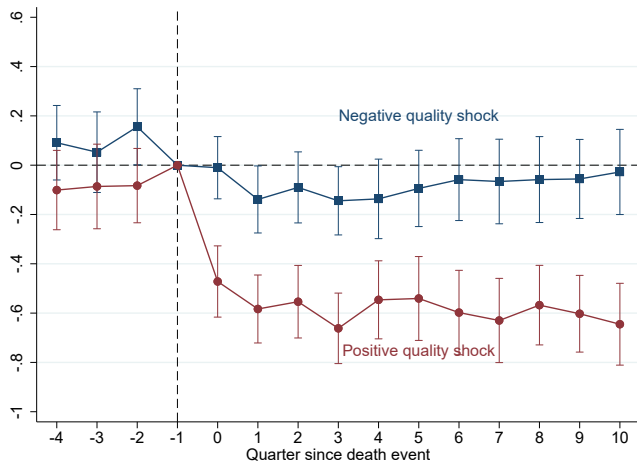


All physicians: $\hat{\beta} = \hat{\gamma} - 1$
(pass-through)

Only **alive** physicians: $\hat{\gamma}$
(endogenous response)

Positive versus negative quality shocks

Figure: Effects of Quality Shocks on Quality Composition of Physicians
(Negative versus Positive Quality Shocks)



Summary of mechanisms

- Quality shocks have no effects on patient volume, so results indeed reflect reallocation
- Reallocation is **almost entirely internal** (even in local labor markets favoring hiring)

► Results

- Response seems to strengthen with **knowledge about the worker**
 - $\hat{\beta} = -0.75$ if the deceased had < 1 year of tenure and $\hat{\beta} = -0.35$ if had $+3$ years

► Results

- Negative shocks are fully mitigated in hospitals with established **formal quality-control practices**, but not in hospitals with limited practices

► Results

- Negative shocks are fully mitigated in hospitals where the baseline **caseload per physician is below the median**, but not in hospitals with high caseload

► Results

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How informative is the observed variation in survival rates about true value added?

- Model relating realized survival y_{hst} to observational physician effects $\bar{\theta}_{hst}$:

$$y_{hst} = \lambda \bar{\theta}_{hst} + \mathbf{FE}_{hs} + \mathbf{FE}_{ts} + \mathbf{Z}'_{hst} \Phi + \bar{\eta}_{hst} + \epsilon_{hst} \quad (3)$$

- ▶ λ is the reliability coefficient and η is the forecast error in $\mu_{hst} = \lambda \theta_{hst} + \eta_{hst}$
 - ▶ $\mathbf{Z}'$ is a vector of risk adjusters and controls
- We use **death-induced quality shocks** as an **IV** to the average observational value-added of physicians

How informative is the observed variation in survival rates about true value added?

Table: Effects on Realized Survival

	Reduced-Form		IV	
	(1)	(2)	(3)	(4)
Observational survival			1.133*** [0.095]	1.140*** [0.090]
Quality shock $\times$ 1{post event}	-0.432*** [0.050]	-0.434*** [0.048]		
Kleibergen-Paap F-statistics			107.49	107.49
Risk adjustment		✓		✓
Basic controls	✓	✓	✓	✓
Pre-period mean	0.948	0.948	0.948	0.948
Observations	2,470,122	2,470,122	2,415,174	2,415,174

- With $\hat{\lambda} \simeq 1$, the quality-composition measure maps strongly into realized survival → measure has high reliability + endogenous responses to the exit of top performers must have averted deaths

Averted deaths

- To gauge magnitudes, consider a one SD negative quality shock: $\sigma = 0.012$
- Absent any targeted reallocation ($\beta = -1$), survival would fall by:

$$100 \times \hat{\lambda} \times 1 \times \sigma \approx 1.36$$

- For negative shocks we estimate $\hat{\beta}_{\text{neg}} \approx -0.14$, so the realized change is

$$100 \times \hat{\lambda} \times 0.14 \times \sigma \approx 0.19$$

- **Targeted reallocation *averts* a $1.36 - 0.19 \approx 1.17$ percentage points or a 20% increase in mortality**

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Summing up

- We use unexpected physician deaths to compare mechanical quality-composition shocks with realized post-exit quality changes
- Hospitals are not blind to physician quality: they mitigate most negative shocks through targeted reallocation
- This response is asymmetric and operates primarily through incumbent physicians
- The response is stronger when hospitals have more information, greater organizational capacity, and more labor slack
- Because physician quality composition predicts realized survival, endogenous reallocation saves lives

Thank You

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- Holmström, B. (2017). Pay for performance and beyond. *American Economic Review*, 107(7):1753–1777.
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Internal versus external replacement

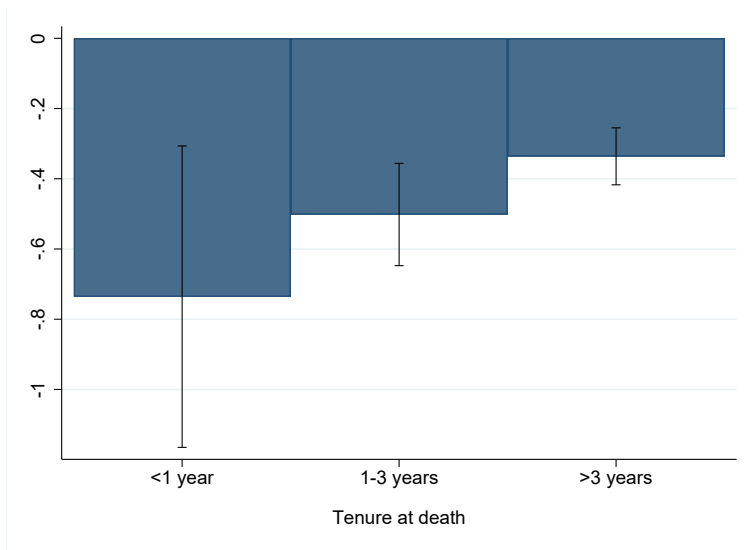
Decompose observable value added in a service **among alive physicians** as the contribution of incumbents and new hires and estimate the effects on each margin ($\hat{\gamma}$)

	Total adjustment (1)	Internal reallocation (2)	External replacements (3)
Quality shock $\times 1\{\text{post event}\}$	0.648*** [0.053]	0.644*** [0.056]	0.006 [0.019]
Basic controls	✓	✓	✓
Pre-period mean	-0.001	-0.001	0
Observations	2,461,168	2,461,168	2,461,168

- The mitigating response is almost entirely internal: hospitals reallocate patients among incumbents rather than replacing the deceased physician externally

[◀ Back to mechanisms summary](#)

Response strengthens with the deceased physician's tenure



Organizational capacity: formal quality-control practices

Formal quality control practices?	Negative quality shocks			Positive quality shocks		
	Established (1)	Partial (2)	Limited (3)	Established (4)	Partial (5)	Limited (6)
Quality shock $\times$ 1{post event}	0.111 [0.191]	-0.151** [0.066]	-0.217** [0.097]	-0.529*** [0.157]	-0.585*** [0.088]	-0.493*** [0.112]
Basic controls	✓	✓	✓	✓	✓	✓
Pre-period mean	0.003	-0.003	-0.002	0.004	-0.003	-0.002
Observations	385,055	1,542,211	445,907	379,365	1,524,446	443,682

- Negative shocks are almost fully mitigated in hospitals with established formal practices, but not in hospitals with limited practices

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Organizational capacity: labor slack

Baseline caseload	Negative quality shocks		Positive quality shocks	
	Low (1)	High (2)	Low (3)	High (4)
Quality shock $\times 1\{\text{post event}\}$	0.012 [0.117]	-0.244*** [0.053]	-0.677*** [0.079]	-0.427*** [0.075]
Basic controls	✓	✓	✓	✓
Pre-period mean	-0.002	0	-0.002	0
Observations	1,194,863	1,178,310	1,177,812	1,169,681

- Negative quality shocks are fully mitigated when baseline workload is low, but generate sizable losses when workload is high
- Effective reallocation requires spare capacity

[◀ Back to mechanisms summary](#)