

# The Role of Human Capital in a Highly-Specialized Context:

## The Case of High-Volume Orthopaedic Surgery

Sahar Paktinat

Department of Economics, University of Bologna, Italy

**PhD Coordinator**  
Prof. Emanuele Padovani

**Supervisor**  
Prof. Rossella Verzulli

**Co-supervisor**  
Prof. Cristina Ugolini

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# Thesis Introduction

- **Orthopaedics** is increasingly central to health systems due to **ageing populations** and **rising demand** for joint replacement.
- Policymakers adopt structural reforms (**centralization, outsourcing, integration**), but their impact is unclear.
- This **dissertation** studies how **organisational arrangements** affect **outcomes**, comparing urgent (**hip fractures**) and elective (**knee arthroplasty**) care.

# Thesis Outline

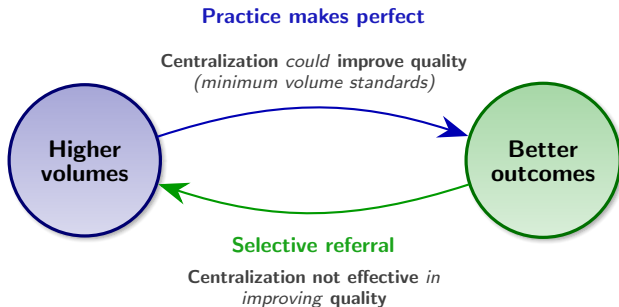
- Chapter I: **"Does Practice Make Perfect? Panel Evidence from Hip Fracture Surgery in Italy"**, joint with Cristina Ugolini, Rossella Verzulli
  - Minimum volume thresholds for emergency hip fractures
- Chapter II: **"When Private Providers Serve the Public Health System: Evidence from an Integrated Healthcare Model"**, joint with Giulia Guidi, Cristina Ugolini, Rossella Verzulli, Barbara Bordini
  - Public and private differences and knee replacement projections
- Chapter III: **"Hospital Volume and Knee Replacement Outcomes: Implications for Workforce Participation"**, joint with Sverre A.C. Kittelsen
  - Labour market outcomes as quality indicators

# Does Practice Make Perfect?

Panel Evidence from Hip Fracture Surgery in Italy

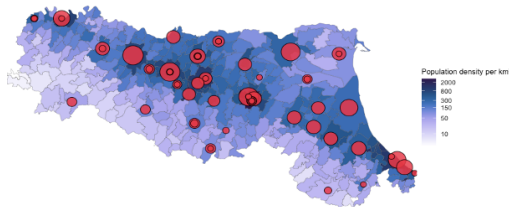
# Introduction

- **Existing literature** shows a **positive association** between the **frequency** a hospital performs a **surgical procedure** and **better patient outcomes** (e.g., Gaynor et al., 2005; Facchini, 2021; Červený, 2023).
- Two possible explanations. **Direction of causality** important for **policy**:



# Context and Setting

- **Highly regulated system:** Emilia-Romagna region
- **Emergency care:** hip fracture surgery within 48h  
→ Patients **cannot choose** hospital or surgeon ⇒ **Selective referral unlikely**
- **Policy:** DM70/2015 imposes **minimum volume threshold** (75 surgeries/year per ward)
- **Research question:** Does higher ward volume reduce **30-day mortality** for hip fracture patients?



*Red circles: wards performing hip fracture surgery; overlap = multiple wards per hospital.*

# Empirical Methodology

- Data: **SDO** linked to the **Mortality register** and the regional surveillance system for surgical site infections (**SICHER**).
- We estimate the effect of surgical experience on patient outcomes using a **multi-way fixed effects model**:

$$y_{iswht} = \beta_0 + \beta_1 \ln(\text{volume})_{iwh,t-1} + X'_{iswht} \beta_2 + \alpha_s + \delta_h + \gamma_t + \varepsilon_{iswht} \quad (1)$$

- Outcome: **30-day mortality** after hip fracture surgery.
- Key variable: **lagged ward-level surgical volume**.
- Controls: patient demographics, comorbidities, fracture type, and weekend admission.
- Fixed effects for **surgeon, hospital, and year** absorb time-invariant heterogeneity.
- $\beta_1$  measures how prior surgical volume affects mortality risk.

# Lagged surgical volumes and 30-day mortality

	(1)		(2)		(3)		(4)	
Ln(Ward volume) <sub>t-4</sub>	-0.00633 (0.00502)	-0.00578 (0.00487)						
Ln(Ward volume) <sub>t-12</sub>			-0.00922 (0.00560)	-0.00643 (0.00545)				
Ln(Ward volume) <sub>t-24</sub>					-0.0212** (0.00915)	-0.0167* (0.00886)		
Ln(Ward volume) <sub>t-52</sub>							-0.0333*** (0.0122)	-0.0291** (0.0113)
Patient controls	N	Y	N	Y	N	Y	N	Y
Surgeon FE	Y	Y	Y	Y	Y	Y	Y	Y
Hospital FE	Y	Y	Y	Y	Y	Y	Y	Y
Year FE	Y	Y	Y	Y	Y	Y	Y	Y
N	10,377	10,377	9,996	9,996	9,327	9,327	7,550	7,550
Bayesian crit. (BIC)	-6373.8	-6684.1	-6173.3	-6468.2	-5508.5	-5804.7	-4429.8	-4660.5
Akaike crit. (AIC)	-6149.1	-6445.0	-5985.8	-6230.2	-5330.0	-5569.1	-4270.5	-4431.9
R <sup>2</sup>	0.0571	0.0853	0.0521	0.0810	0.0574	0.0885	0.0648	0.0953

Notes: Sample period: years 2017–2019. Standard errors (in parentheses) are clustered on hospitals. \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$ .

# Effect of past year hospital ward volume thresholds

	(1)	(2)
<i>Past year hospital ward volume</i>		
(baseline: $\leq 75$ )		
76–118 surgeries	–0.0556 (0.0441)	–0.0558 (0.0395)
119–186 surgeries	–0.0690 (0.0421)	–0.0657* (0.0378)
198–260 surgeries	–0.0776* (0.0432)	–0.0733* (0.0389)
261–285 surgeries	–0.0910** (0.0428)	–0.0849** (0.0384)
286–330 surgeries	–0.0954** (0.0435)	–0.0898** (0.0392)
$\geq 331$ surgeries	–0.103** (0.0461)	–0.0940** (0.0432)
Patient controls	N	Y
Surgeon FE	Y	Y
Hospital FE	Y	Y
Year FE	Y	Y
N	7,550	7,550
$R^2$	0.0640	0.0949
Akaike crit. (AIC)	–4413.7	–4657.3
Bayesian crit. (BIC)	–4219.7	–4428.7

Notes: Sample period: years 2017–2019. Standard errors (in parentheses) are clustered on hospitals. \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$ .

# Conclusions and Policy Implications

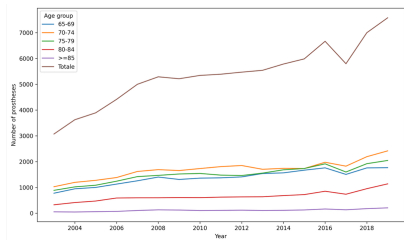
- **Higher** procedural **volumes** are associated with **better outcomes**.
- Results are **consistent** with the **learning-by-doing** dynamic.
- Current **minimum of 75 procedure** per year may **fall short** of fully capturing potential quality gains.
- **Centralisation** may **improve quality** but could **limit access**, adjusting thresholds in line with **local healthcare capacity** and **geographic conditions**.

# When Private Providers Serve the Public Health System:

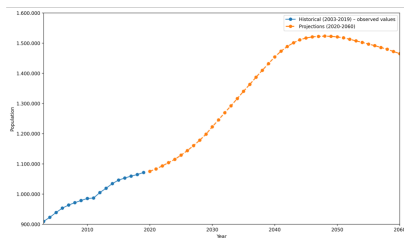
## Evidence from an Integrated Healthcare Model

# Motivation and Research Question

- **Demand for elective orthopaedic surgery is rising** due to ageing and postponed care (COVID-19).
- Health systems increasingly rely on **accredited private providers**.
- **Evidence on quality differences remains mixed**.
- **Question:** Can private provision expand capacity without compromising **long-term quality**?



(a) Knee procedures by age group



(b) Population projections

# Setting & Study design

- **Accredited private hospitals** deliver **publicly funded elective care**.
- **Emilia-Romagna model:** coordinated, planning-based system with strong integration between providers.
- **Study 1 - Survival analysis**
  - Register of Prosthetic Orthopaedic Implants (**RIPO**), 2001–2021.
  - Compare prosthesis survival in public vs accredited private hospitals.
  - Implant failure is defined as **any revision** surgery.
- **Study 2 - Demand forecasting**
  - Use administrative data (SDO) + Eurostat demographic projections.
  - Model and forecast annual volumes of knee arthroplasties in Emilia-Romagna, 2003–2060.
- Together: quality comparison + future demand = framework for **strategic planning** in a highly-integrated system.

# Study 1: Survival Analysis

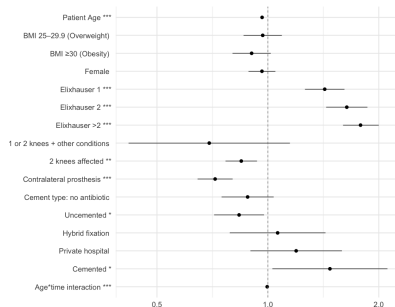
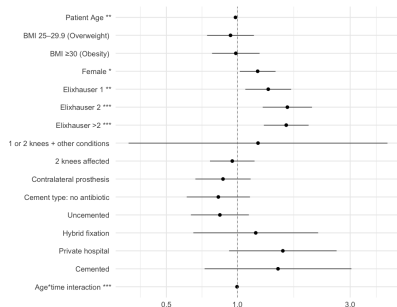
- 1 **Kaplan–Meier survival curves**
- 2 **Cox proportional hazards model** with age as a time-varying covariate (stratified by surgeon)

$$h_{s_i}(t) = h_{0s_i}(t) \exp(Z_i\beta + \beta_{\text{age}} \text{Age}_i + \gamma_{\text{age}} \text{Age}_i \ln(t))$$

## Where:

- $s$  = surgeon
- $h_{0s}(t)$  = surgeon-specific baseline hazard
- $Z_i$  = vector of patient and implant covariates

# Study 1: Cox Results



(a) Cox model hazard ratios for UKA, accounting for a time-varying effect of patient age and stratified by surgeon.

(b) Cox model hazard ratios for TKA, accounting for a time-varying effect of patient age and stratified by surgeon.

# Study 2: Demand Forecasting

## ● Step 1: Count model (NGB1 GLM with log link)

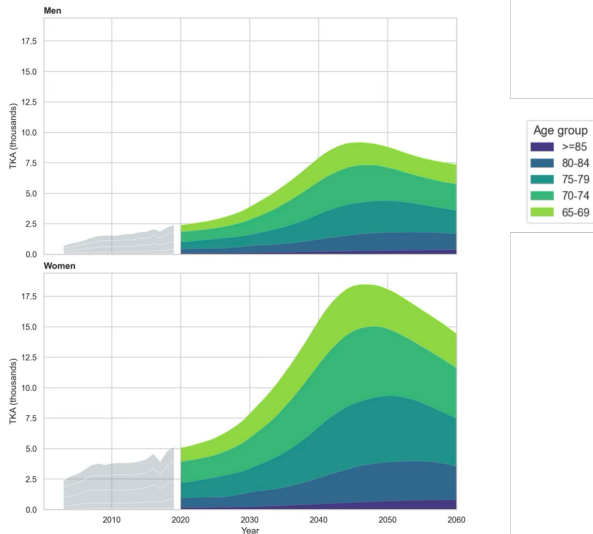
$$\log(\mu_{it}) = \beta_0 + f(\text{year}) + \text{agegroup}_{it} + \text{sex}_{it} + \log(\text{age65}_{it}) + f(\text{year}) \times \text{agegroup}_{it} + \log(\text{pop}_{it}) \quad (2)$$

- $\mu_{it}$ : Expected number of knee replacements in cell  $i$  (sex  $\times$  age group) in year  $t$ .
- $f_k(\text{year}_{it})$ : Spline terms capturing a non-linear time trend.
- $\text{agegroup}_{it}, \text{sex}_{it}$ : Age-group and sex effects.
- $\log(\text{age65}_{it})$ : Effect of population ageing (share aged 65+).
- $f_k(\text{year}_{it}) \times \text{agegroup}_{it}$ : Interactions allowing different time trends by age group.
- $\log(\text{pop}_{it})$ : Offset term controlling for population size.

## ● Step 2: Time-series smoothing of residuals

- Pearson residuals from the NB model are smoothed using an **ETS process**
- Final forecast combines **NB predictions** and **ETS residuals**

## Study 2: Forecasting Results



## Conclusions and Policy Implications

- **No evidence of lower implant survival** in accredited private hospitals after adjustment.
- Knee arthroplasty demand is projected to **grow substantially**, peaking around **2045 (~27,500 procedures)**.
- Expanding accredited private provision could **increase capacity** and **reduce pressure on public hospitals**.

# Hospital Volume and Knee Replacement Outcomes: Implication for Workforce Participation



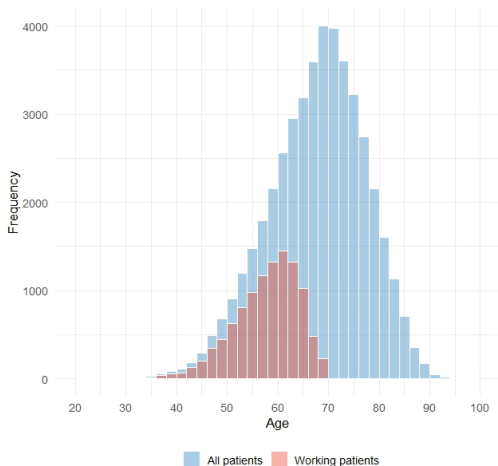
# Introduction

- **Existing literature** documents a **positive association** between **hospital volume** and **patient outcomes**, but the **causal interpretation** remains unclear due to **endogeneity** (e.g., selective referral vs. learning-by-doing).
- **Why this setting?**
  - **Elective procedure** → strong patient choice
  - **Endogeneity concerns more relevant** than in emergency care
  - Allows analysis of both **clinical** and **labour market outcomes**
- **Research question:** What is the causal effect of **hospital volume** on **clinical** and **labour market outcomes** after primary **knee replacement** in Norway?

# Quality indicators

- Two types of **quality indicators**:
  - **Clinical** outcomes:
    - **30-day emergency readmissions**: reflects acute postoperative complications.
    - **revision within 2 years** after **knee replacement** surgery: early implant failure.
  - **Labour market** outcomes:
    - Patient **sick-leave**: Captures **speed of reintegration** into the labour market.
    - **Differential net income**: Reflect ability to sustain employment and earnings.
  - Why labour market outcomes?
    - **Rising procedure rates**.
    - **Return-to-work** as a complementary indicator of surgical success.
    - Employment and income as central to **welfare** → **Societal implications**.

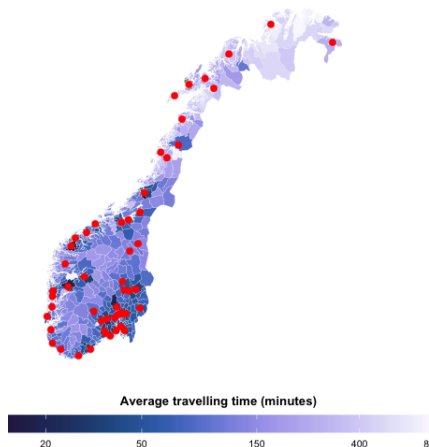
# Patient Descriptives



	Total	Working
Mean age	69	58
Mean travel time	88	101
Median travel time	34	38
Chose closest hospital	47%	44%
Charlson 0	92%	95%
<b>Education level:</b>		
Primary & lower secondary school	25%	16%
Technical college	2%	4%
Upper secondary school	49%	49%
Short university	18%	25%
Long university	5%	6%
<b>Clinical outcomes:</b>		
30-day readmissions	11%	–
Revision 2 years	2%	–
<b>Labour outcomes:</b>		
Sick leave (days)	–	157
Salary pre-surg. (NOK)	–	526,707
Salary post-surg. (NOK)	–	467,295
Diff. net income (NOK)	–	-59,411
Diff. median (NOK)	–	-10,950
Manual labour	–	34%
Obs	43,813	10,731

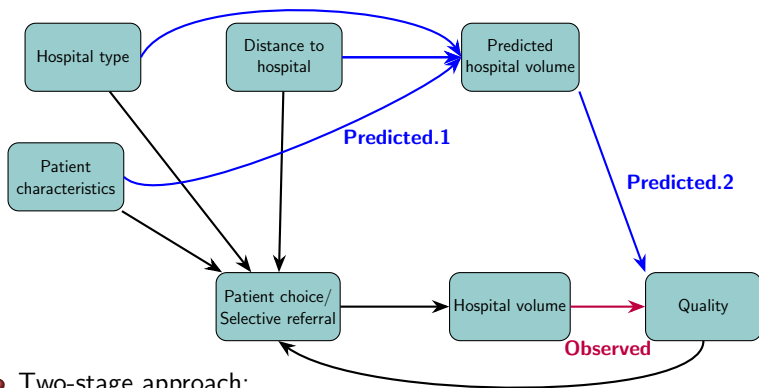
- The exchange rate in 2019 was approximately 10NOK=1EUR

# Travel time



- Data: **NPR** and **SSB**.
- Years of analysis: 2012-2019.
- **Travel time distance:** Distance matrix linking each **patient's municipality** to all **hospitals** in the dataset.
- Adapted code from Asheim (2021) using **Statens vegvesen's routing** system.
- Used their **Routing API** to calculate realistic travel times on the road network.

# Methodology



- Two-stage approach:

- **First stage:** obtain **predicted hospital volume** using a **conditional logit model** of hospital choice.
- **Second stage:** pooled OLS regression of outcomes on observed and predicted hospital volume.

## First stage

- **Conditional logit** model to analyse patient choice of hospital
- Where  $V_{iht}$  is the utility of patient  $i$  derived from hospital characteristics  $h$  at time  $t$ :

$$V_{iht} = \gamma_1 d_{iht} + \gamma_2 d_{iht}^2 + \gamma_3 d_{iht}^3 + \gamma_4 \text{close}_{iht} + z'_{ht} \gamma_z + \sum_{k=1}^K x_{ikt} (\gamma_{1k} d_{iht} + \gamma_{2k} d_{iht}^2 + \gamma_{3k} d_{iht}^3)$$

- $d_{iht}$  represents the **distance** (in terms of **minutes**) between patient  $i$  (home municipality administrative center) and hospital  $h$  (postcode) at time  $t$ .
- **Quadratic** and **cubic** terms of distance to allow for a non-linear effect.
- $\text{close}_{iht}$ , to capture the utility of avoiding any excess travel past the closest hospital.
- $z_{ht}$ : dummy variables for hospital characteristics (i.e. university hospital, private hospital, public hospital) at time  $t$ .

## First stage

- We can then calculate the predicted volume of hospital  $h$  equal to the sum of the estimated probabilities  $\hat{p}_{iht}$  across all patients of choosing hospital  $h$ :

$$\hat{v}ol_{ht} = \sum_{i=1}^N \hat{p}_{iht} = \sum_{i=1}^N \frac{\exp(\hat{V}_{iht})}{\sum_{h' \in M_{it}} \exp(\hat{V}_{iht'})}$$

- The conditional logit model is **estimated separately** for **each year** to account for **changes** in the **set of active hospitals** over time.
- This approach allows **patients' hospital choice preferences to vary by year** instead of assuming constant preference parameters.
- **Predicted hospital volume** is constructed from first-stage conditional logit models and defined as the sum of predicted volumes for clinical and labour market populations.

## Second stage

- Baseline model is a **pooled OLS**:

$$y_{iht} = \beta_0 + \beta_1 \widehat{\text{Volume}}_{ht} + X'_{iht} \beta_2 + \gamma_t + \epsilon_{iht}$$

- $\widehat{\text{Volume}}_{ht}$  represents either the actual volume or the predicted volume of knee replacement surgeries in hospital  $h$  in year  $t$ .
- $X'_{iht}$  patient-level controls (sex, age, number of Charlson comorbidities, patient education level, and — for the labour market outcomes — an indicator for manual occupation).
- $\gamma_t$  year dummies.
- Where **patient outcomes**:
  - Two labour market: **sick leave duration** and **income change**.
  - Two clinical: **30-day readmission** and **2-year revision surgery**.

## Effect of Volume on Clinical Outcomes

30-day readmissions	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	-0.00007** (0.00003)		-0.00000 (0.00006)	
Log(volume)		-0.02891*** (0.00831)		-0.00154 (0.01271)
$R^2$	0.014	0.016	0.012	0.012
Observations	43,913	43,913	43,913	43,913
Revision within two years	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	-0.00001** (0.00000)		0.00002 (0.00002)	
Log(volume)		-0.00204 (0.00179)		0.00358 (0.00347)
$R^2$	0.003	0.003	0.003	0.003
Observations	43,913	43,913	43,913	43,913

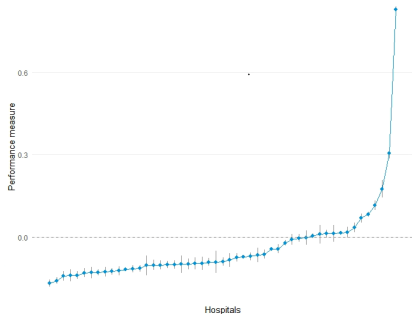
Notes: In parentheses, robust standard errors are clustered at the hospital level. \* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ .

## Effect of Volume on Labour Outcomes

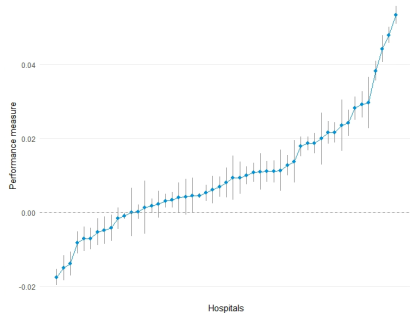
<b>Sick leave</b>	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	0.008 (0.006)		-0.014 (0.024)	
Log(volume)		1.686 (2.164)		-2.971 (4.603)
$R^2$	0.062	0.062	0.062	0.062
Observations	10,731	10,731	10,731	10,731
<b>Differential net income</b>	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	14.97*** (5.14)		16.70 (27.51)	
Log(volume)		5,264.21** (2,092.2)		1,018.92 (5,206.72)
$R^2$	0.031	0.031	0.030	0.030
Observations	10,731	10,731	10,731	10,731

*Notes:* In parentheses, robust standard errors are clustered at the hospital level. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

# Variation in Clinical Outcome Quality Indicators

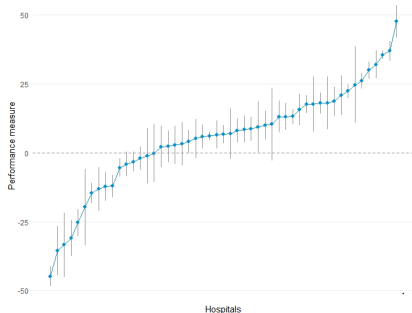


(a) 30-day emergency readmission

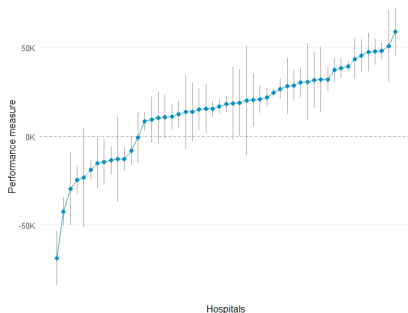


(b) Revision within two years

# Variation in Labour Market Quality Indicators



(a) Patient sick leave



(b) Differential net income

## Conclusions & Policy implications

- When accounting for reverse causality using **predicted hospital volume** we find **no** effect on quality.
- Since volume **does not** improve outcomes:
  - We **cannot support centralization** of elective procedures like knee replacements.
  - We should still support policies to **increase** patient access to high-quality, **high-volume** hospitals.
- Attention should shift to **other drivers** of hospital performance: **surgeon-level experience, organisational practices, and resource allocation.**
- Policies promoting **patient and GP** information on **hospital quality** indicators should be supported.

# Thesis Conclusion

- Does organisation shape orthopaedic outcomes? *It depends on the procedure.*

	Hip Fracture (Ch.1)	Knee Arthroplasty (Ch.2)	Knee Arthroplasty (Ch.3)
Setting	Emilia-Romagna	Emilia-Romagna	Norway
Key Finding	Higher volume → lower mortality	Public = Private quality	Volume → no effect on clinical/labour market outcomes
Policy Implication	Centralisation justified	Outsourcing viable	Centralisation not justified

## Three Takeaways:

- 1 **Volume matters — but only for urgent procedures** where experience and coordination are decisive
- 2 **Outsourcing routine procedures** can relieve public sector pressure without quality loss
- 3 **Centralisation carries equity costs** only worth bearing when outcome gains are demonstrated

*Future research should look beyond hospital volume toward surgeon-level experience and organisational practices as drivers of performance.*

# Thank you for your attention!

sahar.paktinat2@unibo.it

LinkedIn:



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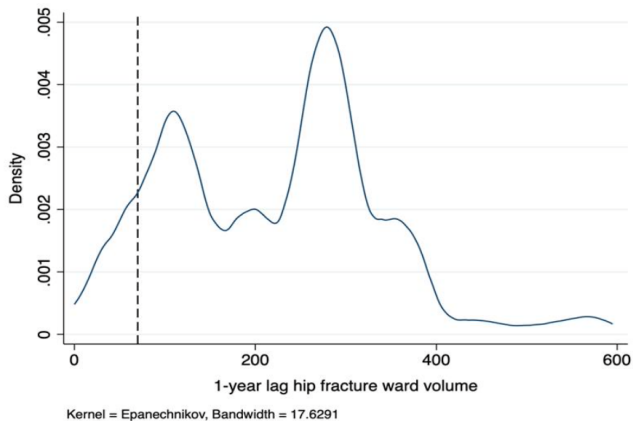
Chapter III: Robustness Checks

# Chapter I: Patient summary statistics

	Mean	Std. Dev.	Min	Max
<i>Patient characteristics</i>				
30-day mortality	0.034	0.183	0	1
Age	83.90	7.437	65	100
Female	0.755	0.430	0	1
Pre-operative wait > 48 hours	0.205	0.404	0	1
Post-surgical length of stay	9.896	6.244	0	175
CCI 0	0.659	0.474	0	1
CCI 1	0.195	0.396	0	1
CCI 2	0.081	0.274	0	1
CCI over 2	0.064	0.245	0	1
ASA I	0.010	0.101	0	1
ASA II	0.233	0.423	0	1
ASA III	0.646	0.478	0	1
ASA IV	0.109	0.312	0	1
ASA V	0.001	0.034	0	1
Admission during weekend	0.272	0.445	0	1
Operation during weekend	0.124	0.330	0	1
<i>Type of surgery</i>				
Partial or total hip replacement	0.497	0.500	0	1
Fracture reduction	0.502	0.500	0	1
<i>Type of hip fracture</i>				
Petrochanteric	0.438	0.496	0	1
Transcervical	0.535	0.499	0	1
Other type of fracture	0.027	0.163	0	1
Observations	10,377			
<i>Ward volumes for hip fracture surgeries</i>				
Surgeries in the past 4 weeks	17.825	0.325	0	60
Surgeries in the past 12 weeks	53.149	28.985	0	158
Surgeries in the past 24 weeks	105.260	56.369	0	298
Surgeries in the past 52 weeks	224.453	116.918	0	595

*Notes:* Sample period: years 2017–2019. ASA: American Society of Anesthesiologists.

# Chapter I: Kernel density



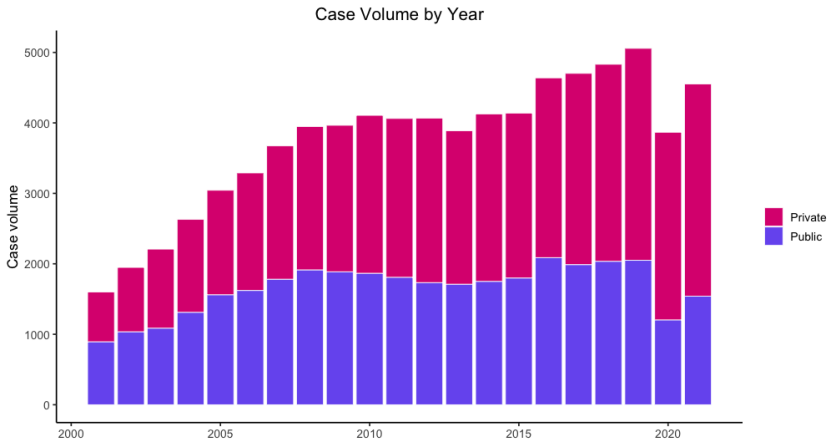
## Chapter II: Patient Characteristics

**Table 2.2:** Patient and procedure characteristics of all primary total knee arthroplasties (TKA) and unicompartmental knee arthroplasties (UKA), by type of hospital, years 2001–2021.

Variable	TKA			UKA		
	Private	Public	Diff t-stat	Private	Public	Diff t-stat
	<i>n</i> = 38,676 Mean SD	<i>n</i> = 31,355 Mean SD	(95% CI)	<i>n</i> = 5,233 Mean SD	<i>n</i> = 3,602 Mean SD	(95% CI)
Age < 60	.087 (.281)	.068 (.252)	-.019 (-.023 to -.015)***	.221 (.415)	.160 (.367)	-.061 (-.078 to -.044)***
Age 60–74	.534 (.499)	.532 (.499)	-.002 (-.009 to .005)	.572 (.495)	.603 (.489)	.031 (.010 to .052)***
Age ≥ 75	.379 (.485)	.400 (.490)	.021 (.013 to .028)***	.206 (.404)	.236 (.425)	.030 (.013 to .048)***
Female sex	.714 (.452)	.693 (.461)	-.021 (-.027 to -.014)***	.643 (.479)	.679 (.467)	.036 (.015 to .056)***
<i>Fixation</i>						
Cemented	.951 (.215)	.937 (.242)	-.013 (-.017 to .010)	.929 (.255)	.954 (.209)	-.024 (-.014 to .034)***
Cementless	.038 (.193)	.038 (.192)	.000 (-.002 to .003)	.048 (.214)	.041 (.201)	-.006 (-.015 to .002)**
Hybrid	.009 (.097)	.021 (.146)	.011 (.009 to .013)***	.017 (.050)	.032 (.131)	-.015 (-.019 to .010)***
<i>BMI category</i>						
BMI < 18.5	.002 (.047)	.002 (.041)	-.000 (-.001 to .000)*	.001 (.031)	.002 (.044)	-.006 (-.000 to .002)
BMI 18.5–25	.137 (.344)	.153 (.360)	.015 (.010 to .021)***	.171 (.377)	.164 (.370)	-.007 (-.023 to .009)
BMI 25–30	.356 (.480)	.389 (.488)	.033 (.026 to .040)***	.385 (.486)	.423 (.494)	.038 (.017 to .059)***
BMI ≥ 30	.504 (.499)	.456 (.498)	-.048 (-.056 to -.041)***	.442 (.496)	.410 (.492)	-.032 (-.053 to -.010)***
<i>Elixhauser indices</i>						
Elixhauser 0	.316 (.002)	.216 (.002)	-.100 (-.107 to -.094)***	.427 (.007)	.288 (.007)	-.139 (-.160 to -.119)***
Elixhauser 1	.194 (.396)	.190 (.392)	-.004 (-.010 to .002)	.218 (.006)	.221 (.007)	.002 (-.015 to .019)
Elixhauser 2	.151 (.001)	.162 (.002)	.011 (.005 to .016)***	.135 (.005)	.168 (.006)	.033 (.018 to .048)***
Elixhauser >2	.338 (.002)	.432 (.003)	.093 (.086 to .100)***	.218 (.006)	.323 (.008)	.104 (.086 to .123)***

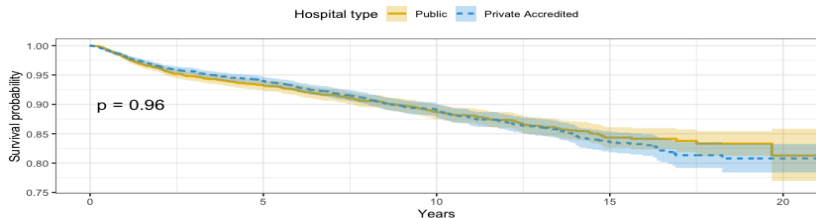
Notes: \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

## Chapter II: Knee Replacement Volume



**Figure 5:** Annual case volume by sector, showing public and private contributions to total cases over time.

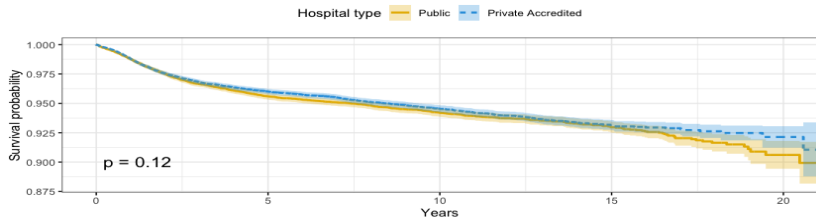
# Chapter II: Kaplan-Meier UKA Results



Group	0 Years	5 Years	7 Years	10 Years	15 Years	20 Years
Public	100%	93.2%	91.6%	88.7%	84.3%	81.3%
Private Accredited	100%	93.9%	92%	88.9%	83.6%	80.8%

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# Chapter II: Kaplan-Meier TKA Results



Group	0 Years	5 Years	7 Years	10 Years	15 Years	20 Years
Public	100%	95.6%	95.1%	94.2%	93%	90.6%
Private Accredited	100%	96%	95.5%	94.5%	93.1%	92.1%

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## Chapter II: Model choice

- Model choice: NB1 preferred to NB2 (lower AICc, better Pearson dispersion).
- Fixed holdout (2017–2019):
  - NB1 baseline: MAPE 22.8%
  - NB1 + ETS: MAPE 10.0%

**Table 1:** Goodness of fit comparisons on predictive accuracy.

Model	Validation Period	MAPE (%)
NB1 (baseline)	2017–2019	22.77
NB1 + ETS (selected)	2017–2019	10.03
Log-linear (Romanini et al.)	2017–2019	100.00

- **MAPE** = mean absolute percentage error

## Chapter II: Model choice

- 8 MAPE comparison across three validation windows

Table 2: Robustness on different holdouts.

Window	MAPE NB1 (%)	MAPE NB1 + ETS (%)
2014–2016	19.7	18.7
2015–2017	21.1	18.4
2016–2018	24.8	23.4

- NB1 + ETS consistently outperforms NB1 alone.

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## Chapter II: Empirical Strategy Identification

- **NB1** specification provides the best goodness of fit to our historical data.

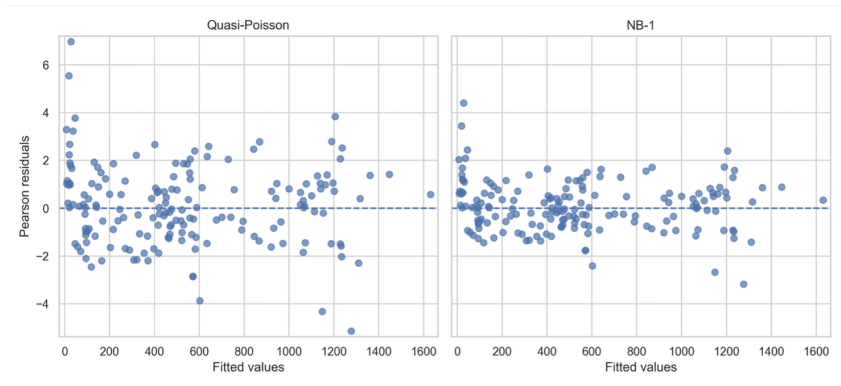


Figure 6: Fitted values vs Pearson residuals.

# Chapter II: Empirical Strategy Identification

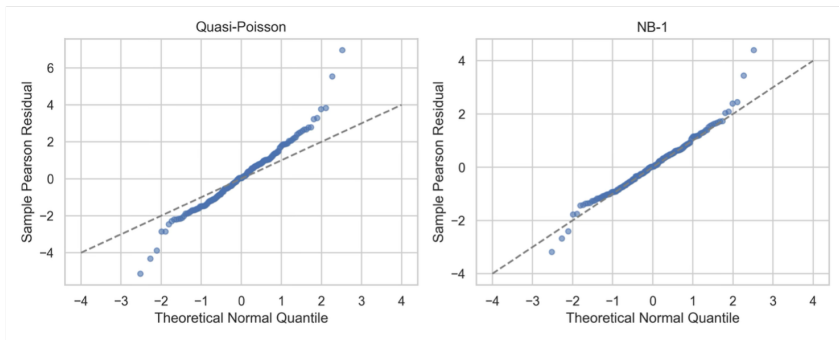


Figure 7: Q-Q plots.

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## Chapter II: Goodness of fit

**Table 3:** Goodness of fit comparisons of the Negative Binomial Models.

<b>Model</b>	$\alpha$	<b>Pearson</b>	<b>AICc</b>
NB1	1.5829	1.25	1687.6
NB2	0.00026	1.78	1723.8

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## Second stage model explanation

- **Predicted hospital volume** is constructed from first-stage conditional logit models and defined as the sum of predicted volumes for clinical and labour market populations.

$$\text{Volume}_{ht} = \text{Volume}_{ht}^{\text{CL}} + \text{Volume}_{ht}^{\text{LM}}$$

- Where **patient outcomes**:
  - Two labour market: **sick leave duration** and **income change**.
  - Two clinical: **30-day readmission** and **2-year revision surgery**.
- Second-stage regressions are run separately by outcome and relevant population.

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# Effect of volume on 30-day emergency readmission

Panel A: Pooled OLS	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	-0.00007** (0.00003)		-0.00000 (0.00006)	
Log(volume)		-0.02891*** (0.00831)		-0.00154 (0.01271)
$R^2$	0.014	0.016	0.012	0.012
Observations	43,913	43,913	43,913	43,913
Panel B: Inclusion of Hospital FE	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	-0.00005 (0.00004)		-0.00007 (0.00008)	
Log(volume)		-0.02414** (0.01023)		-0.03682* (0.01985)
$R^2$	0.041	0.042	0.041	0.041
Observations	43,913	43,913	43,913	43,913

*Notes:* In parentheses, robust standard errors are clustered at the hospital level. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

# Effect of volume on patient revision within two years

Panel A: Pooled OLS	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	-0.00001** (0.00000)		0.00002 (0.00002)	
Log(volume)		-0.00204 (0.00179)		0.00358 (0.00347)
$R^2$	0.003	0.003	0.003	0.003
Observations	43,913	43,913	43,913	43,913
Panel B: Inclusion of Hospital FE	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	-0.00000 (0.00001)		0.00003* (0.00002)	
Log(volume)		-0.00093 (0.00286)		0.01121** (0.00513)
$R^2$	0.009	0.009	0.009	0.009
Observations	43,913	43,913	43,913	43,913

Notes: In parentheses, robust standard errors are clustered at the hospital level. \* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ .

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# Effect of Volume on sick leave

Panel A: Pooled OLS	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	0.008 (0.006)		-0.014 (0.024)	
Log(volume)		1.686 (2.164)		-2.971 (4.603)
$R^2$	0.062	0.062	0.062	0.062
Observations	10,731	10,731	10,731	10,731
Panel B: Inclusion of Hospital FE	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	-0.015 (0.014)		-0.001 (0.033)	
Log(volume)		-0.724 (3.468)		3.549 (9.041)
$R^2$	0.075	0.075	0.075	0.075
Observations	10,731	10,731	10,731	10,731

*Notes:* In parentheses, robust standard errors are clustered at the hospital level. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

# Effect of volume on differential net income

Panel A: Pooled OLS	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	14.97*** (5.14)		16.70 (27.51)	
Log(volume)		5,264.21** (2,092.2)		1,018.92 (5,206.72)
$R^2$	0.031	0.031	0.030	0.030
Observations	10,731	10,731	10,731	10,731
Panel B: Inclusion of Hospital FE	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	26.51 (28.74)		59.17 (81.28)	
Log(volume)		5,913.26 (7,703.62)		10,152.32 (19,906.81)
$R^2$	0.033	0.033	0.033	0.033
Observations	10,731	10,731	10,731	10,731

Notes: In parentheses, robust standard errors are clustered at the hospital level. All monetary values are expressed in NOK. \* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ .

# Robustness Checks - 30-day emergency readmission

Panel A: Pooled OLS	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	-0.00012** (0.00055)		0.00000 (0.00009)	
Log(volume)		-0.02929*** (0.00853)		-0.01187 (0.012698)
$R^2$	0.014	0.016	0.011	0.012
Observations	43,913	43,913	43,913	43,913
Panel B: Inclusion of Hospital FE	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	-0.000085 (0.00007)		-0.00011 (0.00013)	
Log(volume)		-0.02547*** (0.01015)		-0.03500* (0.01934)
$R^2$	0.041	0.042	0.040	0.040
Observations	43,913	43,913	43,913	43,913

Notes: In parentheses, robust standard errors are clustered at the hospital level. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

# Robustness Checks - two-year revision

Panel A: Pooled OLS	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	-0.00002** (0.00006)		0.000026 (0.000028)	
Log(volume)		-0.00223 (0.00179)		0.00323 (0.00334)
$R^2$	0.003	0.003	0.003	0.003
Observations	43,913	43,913	43,913	43,913
Panel B: Inclusion of Hospital FE	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	-0.00001 (0.00002)		0.00005** (0.00003)	
Log(volume)		-0.00107 (0.00298)		0.01127** (0.00495)
$R^2$	0.009	0.009	0.009	0.010
Observations	43,913	43,913	43,913	43,913

*Notes:* In parentheses, robust standard errors are clustered at the hospital level. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

# Robustness Checks - Sick leave

Panel A: Pooled OLS	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	0.025 (0.016)		-0.034 (0.0651)	
Log(volume)		1.877 (2.209)		-2.645 (4.611)
$R^2$	0.062	0.062	0.062	0.062
Observations	10,731	10,731	10,731	10,731
Panel B: Inclusion of Hospital FE	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	-0.043 (0.040)		-0.004 (0.093)	
Log(volume)		-1.329 (3.442)		1.809 (8.271)
$R^2$	0.075	0.075	0.075	0.075
Observations	10,731	10,731	10,731	10,731

Notes: In parentheses, robust standard errors are clustered at the hospital level. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

# Robustness Checks - Differential net income

Panel A: Pooled OLS	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	44.52** (13.70)		55.44 (74.39)	
Log(volume)		5,660.73** (1,996.38)		2,070.02 (5,280.31)
$R^2$	0.028	0.028	0.028	0.028
Observations	10,731	10,731	10,731	10,731
Panel B: Inclusion of Hospital FE	Observed hospital volumes		Predicted hospital volumes	
	(1)	(2)	(3)	(4)
Continuous volume	49.78 (78.72)		104.26 (223.41)	
Log(volume)		4,855.85 (7,457.85)		5,415.32 (19,294.66)
$R^2$	0.031	0.031	0.029	0.029
Observations	10,731	10,731	10,731	10,731

Notes: In parentheses, robust standard errors are clustered at the hospital level. All monetary values are expressed in NOK. \* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ .