

Dynamic Bargaining between Hospitals and Insurers

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April 22, 2026

Medicare Policy's Overlooked Effect on Private Spend

- Medicare rates fall relative to hospital reported costs (AHA, 2022)
- \$400b+ in annual private spend uses benchmarks like Medicare
 - E.g. “insurer + patient will pay 150% of amount Medicare would pay”
- Turns out, benchmark multiples are in place for multiple years

How would private insurer spending have changed if Medicare benchmark rates went up faster?

Multiyear Contracts Mean Medicare Dynamics Do Something

- Conventional wisdom: benchmark levels don't matter
 - If Medicare pays \$100, negotiate 150% of Medicare \Rightarrow get \$150 payment
 - If Medicare pays \$300, negotiate 50% of Medicare \Rightarrow same dollar payment
- Fixed markups for 3+ years \Rightarrow ex post, Medicare rates directly shape private rates
- But forward-looking negotiators can respond to anticipated change
 - Intuition: surprise effects, but long-run NPV neutrality
 - That intuition \approx holds for one contract, but not the market

Forward-Looking Response to Future Increase

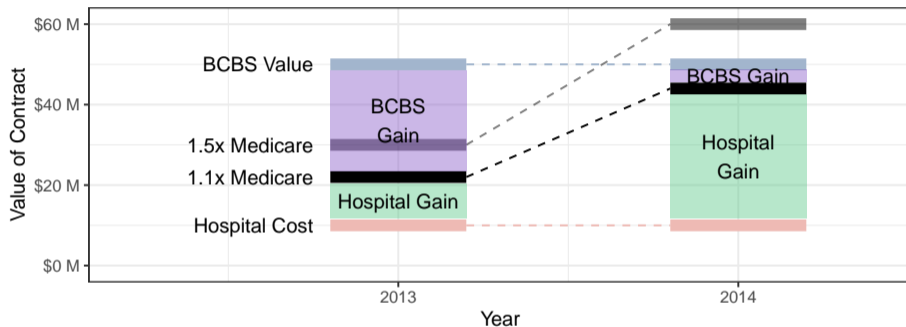


Figure: Illustration of forward-looking response: Large General Hospital negotiates a two-year Medicare-benchmarked contract with BCBS in 2013. The response to a future benchmark-driven payment increase (Δ) is a starting payment decrease ($\beta\Delta$).

Staggering + Discounting Breaks Market NPV Neutrality

- Two hospitals form two-year contracts, $P_{New} = \$30m - \beta\Delta$, $P_{Old} = \$30m + \Delta$
- **Simultaneous** contracts: path changes, but market NPV neutrality in long run

$$P_{New} + P_{New} + \beta(P_{Old} + P_{Old}) = \$60m(1 + \beta) - 2\beta\Delta + \beta2\Delta$$

- **Staggered** contracts: no NPV neutrality in long run

$$P_{New} + P_{Old} + \beta(P_{Old} + P_{New}) = \$60m(1 + \beta) + \Delta(1 - \beta^2)$$

- Advanced announcements add subtlety, but same sign from competitive interactions
- So, are contracts multiyear & staggered? Then, do negotiators time-discount?

Preview: Medicare Has Real Effects on Private Spend

1. Framework (just now): multiyear, stagger, & $\beta < 1 \Rightarrow$ LR effects
 - One-year contracts \Rightarrow no effect on payments
 - Simultaneous contracts \Rightarrow long-run **neutral** in NPV terms
 - Staggering + time discounting \Rightarrow no long-run NPV neutrality
2. Data: contracts are multiyear & staggered
 - Public record data on hospital–insurer contracts from West Virginia
 - One “contract:” an agreement from start to end
 - Potential for real effects: agreements 3+ years and staggered
3. Structural model: real effects if Medicare tracked reported costs
 - Dynamic extension of Ho and Lee (2017), estimate $\hat{\beta} = 0.899$
 - Static model would miss real effects: year-nine effect extrapolates to \$5b
 - Forward-looking response is first-order: payments in one year go down

- Macro-inflation own NPV neutrality $\not\Rightarrow$ market neutrality: I study bargained prices

Taylor (1980), Calvo (1983), Abbott (1995), Acquatella et al. (2023)

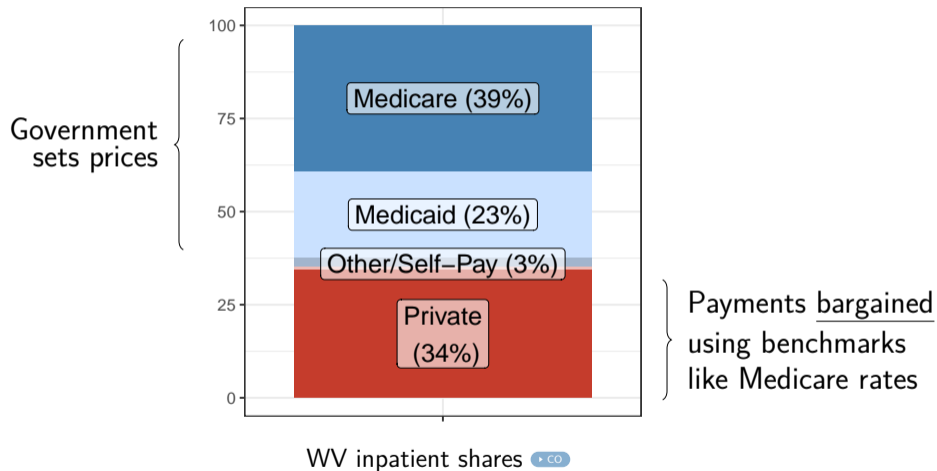
- Benchmarks are associated with market power: I study dynamic implications

Cooper et al. (2019), Weber et al. (2019), Koos et al. (2026)

- Medicare spillovers on private market: I study benchmark dynamics

Clemens and Gottlieb (2017), Clemens et al. (2017), Cooper et al. (2019)

Research Focus: Benchmarks in the Private Market



There Are Two Common Benchmarks

Thousands of services \Rightarrow impractical to negotiate service-by-service

- 1 Medicare-set rates (BCBS, national, my focus)
- 2 Hospital-set list prices (non-BCBS, endogenous, future work)

How did dynamics in these benchmark prices affect real payments in WV?

Different Benchmarks Have Different Dynamic Implications

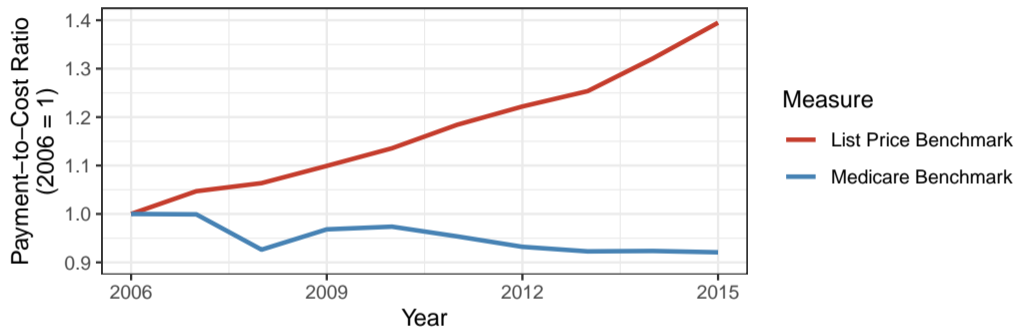


Figure: Medicare payments decreased relative to reported costs while list prices increased quickly.

► Payments Empirical question: what if the blue line was roughly flat instead?

Public record contract report panel

- Data from 2006–15 West Virginia
- Scans of annual hospital reports [▶ Example](#)
- Use for payment rates [▶](#), networks [▶](#), benchmarks , timing, ...
- Other data: inpatient discharges (2016), state-level premiums & sales (2006–16)

Medicare-Based Benchmarks Associated with Bigger Insurers

Insurer	Medicare	List Prices
All	46.74	53.26
Modeled	60.20	39.80
Highmark BCBS	72.27	27.73
HPUOV	56.24	43.76
Other Modeled	13.14	86.86
Nonmodeled	3.03	96.97

Table: Estimated percentage of 2011-16 projected inpatient payments classified as Medicare-benchmarked and list price-benchmarked [▶ Algorithm](#).

Contracts Are Multiyear and Staggered

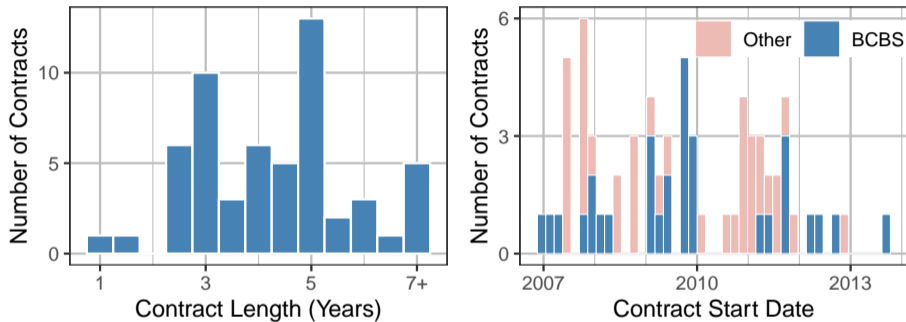




Figure: Contracts were multiyear for BCBS [▶ and others](#) and were staggered, leaving scope for benchmark dynamic effects.

Empirical Model: Ho and Lee (2017) + Dynamics

1. Price benchmarks (Medicare prices and list prices) are updated 
2. Hospitals and insurers simultaneously bargain new contracts
 - Contracts can last more than one period — annual discounting rate β
 - Use Kalai solution (Dorn, 2026) to extend static Nash to dynamic model  More
3. Consumers choose plans and get sick \Rightarrow hospital, insurer demand D^{Hosp}, D^{Ins}
4. Flow profits realized with **price** externalities — some internalized

$$\pi_j^{Ins} = \underbrace{D_j^{Ins}(\cdot)(\phi_j - \eta_j)}_{\text{Premium revenue net of } \eta} - \underbrace{\sum_{h \in \mathcal{G}_j^{Ins}} D_{hj}^{Hosp}(\cdot) p_{hj}}_{\text{Payments to hospitals}} \quad \text{and} \quad \pi_i^{Hosp} = \underbrace{\sum_{n \in \mathcal{G}_i^{Hosp}} D_{in}^{Hosp}(\cdot)(p_{in} - c_i)}_{\text{Payments received - cost of care}}$$

What Do We Need to Estimate?

- Saw from data: contracts are multiyear and staggered
- Main remaining parameter is discounting rate β
 - Larger $\beta \Rightarrow$ bigger forward-looking response \Rightarrow smaller LR effects
 - Bargaining weights (τ) and costs (c, η) govern price levels, but turn out to be less important for dynamic response

Estimation Roadmap

Estimation step	Data inputs	Method outputs	Timing	Identification
1. Hospital demand	BCBS inpatient choice	Hospital network WTP (utils)	Static	Observables logit
2. Insurer demand	H demand, sales estimates, Census	Network effect on sales	Static	ACA area FEs
3. Contract bargaining	Multiple years of payments, demand	Discounting β — this is new	Dynamic	Firm Dummies
4. Benchmark	Prices over time	Observed	Implicit	Observed

Adapted and extended based on Ho (2006)

Step 1: Hospital Demand

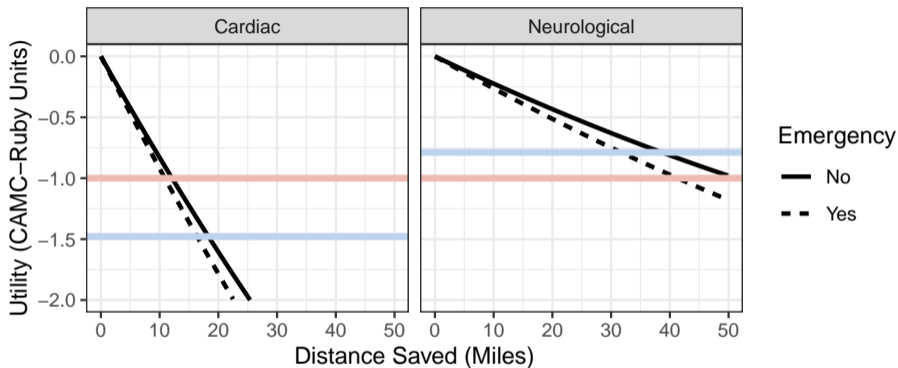


Figure: Specification: $u_{i,h,\ell}^{Hosp} = \delta_{h,\ell}^{Hosp} + \nu_{i,h,\ell}\rho + \varepsilon_{i,h,\ell}$. Figure: estimated diagnosis-distance interactions $\nu_{i,h,\ell}$ for non-emergency (solid) and emergency (dashed) discharges, in CAMC-WVU Ruby units (red line at -1.0). Blue horizontal line is United Hospital-Ruby value. [▶ Table](#)

Step 2: Insurer Demand

$$u_{i,j,c}^{Ins} = \gamma_k WTP_{j,k,c} + \tilde{\delta}_{j,m}^{Ins} + \xi_{j,k,c} + \varepsilon_{i,j,c,m}$$

- Individual i of age group k choosing insurer j in county c in rating area m
- Control for premiums with $\tilde{\delta}_{j,m}^{Ins}$ area FEs
- Calibrate premium sensitivity from Ho (2006)
- Moment $E[WTP_{j,k,c}\xi_{j,k,c}] = 0$ for each age group k [▶ Details](#)

Insurer Demand: Willingness to Pay Coefficients

WTP Coefficient				
γ_{0-17}	γ_{18-44}	γ_{45-64}	γ_{65-74}	γ_{75+}
26.6***	4.94***	2.76***	2.79***	2.05***
(2.65)	(0.67)	(0.33)	(0.27)	(0.15)

Note:

* $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Table: Coefficients on Willingness to Pay parameters. All consumers value network quality, so an insurer that loses a hospital will lose enrollment, giving hospitals leverage.

Step 3: Bargaining Identifies β From Benchmark Variation

- Suppose we see a two-year agreement with $c^H = 10$ & $v^M = 50$ constant
 - Bargain to split NPV gains $(50 - 10) + \beta \times (50 - 10)$ with $\beta = 1/2$ unknown
 - Observe $p_1 = 20$ and $p_2 = 50$; insurer share is known to be $\tau = 1/2$
- GMM recovers β via $E[p_1 - c + \beta(p_2 - c) - (1 - \tau)GFT_1 - \beta(1 - \tau)GFT_2] = 0$
 - $\beta = 0$ implies $1 - \tau = \frac{10+0 \times 40}{40+0 \times 40} = \frac{1}{4} \Rightarrow$ data rejects $\beta = 0$
 - $\beta = 1/2$ implies $1 - \tau = \frac{10+0.5 \times 40}{40+0.5 \times 40} = \frac{30}{60} = \frac{1}{2} \Rightarrow \hat{\beta} = \frac{1}{2}$
- Need $p_2/p_1 - GFT_2/GFT_1$ variation: use list prices vs. Medicare by firm
- Since τ is estimated, identify from restrictions on τ_{ij} heterogeneity

Estimated Gain From Trade Shares (Will Change)

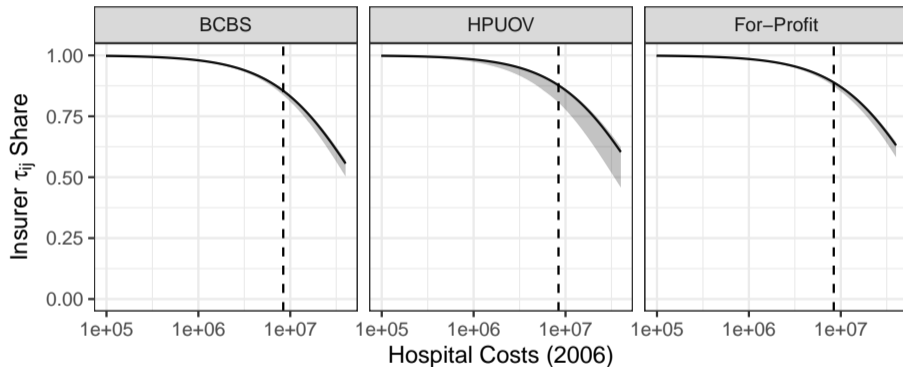


Figure: Share of gains from trade retained by the insurer under estimated dynamic model.

▶ Myopic

I Find Negotiators are Forward-Looking

		Parameter			
	β	τ_{BCBS}	τ_{HPUOV}	τ_{FP}	$-\tau^{Size}$
Myopic	.	0.876***	0.825***	0.861***	1.037***
	(.)	(0.012)	(0.232)	(0.034)	(0.199)
Forward-Looking	0.899***	0.854***	0.877***	0.889***	0.989***
	(0.03)	(0.006)	(0.026)	(0.005)	(0.028)

Note:

* $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

▸ Estimation Details

▸ Other Parameters

▸ Other Models

▸ Moment

Counterfactual Question: Faster Medicare Rate Increase

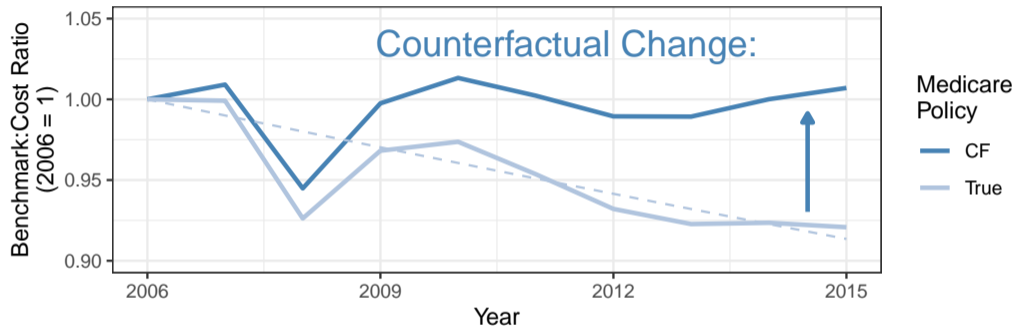


Figure: Medicare payments decreased relative to reported costs by \approx one percentage point annually (dashed line). How would one ppt annual increases have affected spending? [▶ Payments](#)

Counterfactual and Core Mechanisms

- Surprise Medicare announcement at end of 2006
 - Counterfactual Medicare will counteract depreciation relative to hospital reported costs [▶](#)
 - One percentage point annual price increase going forward (relative to actual)
 - Will hold expiration and benchmark choice (+ choice set) fixed [▶ Details](#)
- Conventional static view: no effect
- **Mechanical**: future prices increase
- **Quantify**: starting prices decrease ($\beta > 0$ but $\beta < 1$)

Counterfactual: Meaningful Effects + Meaningful Response

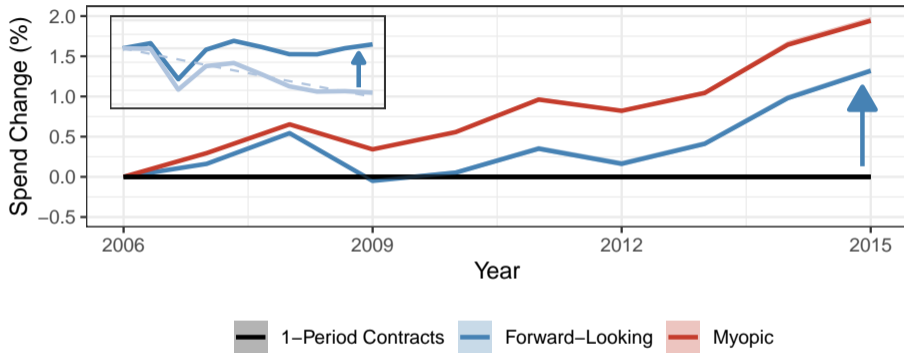


Figure: Counterfactual 2015 payments increase by an estimated 1.319%. Missing dynamic response would overestimate effect on spending by 45%+.

[▶ Decomposition](#) [▶ Timing](#) [▶ By Insurer](#) [▶ By Hospital](#) [▶ \$\beta = 0.97\$](#) [▶ Premiums](#) [▶ Drop CAMC](#) [▶ List Price Caps](#) [▶ Limitations](#) [▶ 45%+ Overestimate](#) [▶ Construction \(Time Series-Adjacent\)](#)

Conclusion

- Benchmark dynamics can have meaningful effects
 - Raw data: contracts are multiyear and staggered
 - Need a dynamic model, or else miss real effects
- But forward-looking responses are first-order
 - Structural model: reject myopia, but $\beta < 1$
 - Ignoring responses leads to effect overestimates of 45%+
- Broader work: when do contract dynamics matter?

Feedback welcome! jacobdorn@cornell.edu

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(Non)Neutrality, Formalized

Proposition 1

Suppose firms 1 and 2 reach two-period agreements to target a time-weighted average payment P^* . Let t_0 be a period in which an agreement is reached, let $k \geq 0$, and let the inflation rate be ϕ . Then if contracting is simultaneous,

$\sum_{t=t_0}^{t_0+1+2k} \beta^t (p_t^{(1)} + p_t^{(2)}) = \sum_{t=t_0}^{t_0+1+2k} \beta^t 2P^*$. But if contracting is alternating, then

$\sum_{t=t_0}^{t_0+1+2k} \beta^t (p_t^{(1)} + p_t^{(2)}) = \sum_{t=t_0}^{t_0+1+2k} \beta^t \left(2 + \frac{(1-\beta)\phi}{1+\beta(1+\phi)} \right) P^*$ 

Example West Virginia Contract Report Scan

Discount Contract List
Budgeted Discounts for FY 2016
Hospital Name Charleston Surgical Hospital

Name of Third Party Payor	Inpatient %	Outpatient %	Inpatient	Outpatient
1 C&O Employees (auto-renewal)	N/A	6 00% ✓	Must Separate	Combine
2 Select-Net (auto-renewal)	10 00% ✓	10 00% ✓	Combine	Combine
3 Cigna (auto-renewal)	18 00% ✓	15 00% ✓	Combine	Combine
4 4Most (auto-renewal)	5 00% ✓	5 00% ✓	Combine	Combine
5 MDI (auto-renewal)	15 00% ✓	10 00% ✓	Combine	Combine
6			Combine	Combine

List discounts in lower section that are (1) new or not currently approved contracts, (2) non-third party (e.g. admin. adj.), (3) contracts with utilization > calculated volume threshold above*, (4) HMO or risk contracts, or, (5) top section of template determined that it must be separated

1 Mt State-PPO	43 38%	41 58%	Must Separate	Must Separate
2 Mt State-Indemnity	43 38%	38 45%	Must Separate	Must Separate
3 Aetna	18 00% ✓	15 00% ✓	Must Separate	Must Separate
4 Carelink	15 00% ✓	13 00% ✓	Must Separate	Must Separate
5 United	10 00% ✓	10 00% ✓	Must Separate	Must Separate

Figure: Charleston Surgical Hospital report, fiscal year 2016. Mountain State/Highmark BCBS generally used Medicare as a benchmark (non-round numbers) while other smaller insurers generally used list prices. [▶ Was WV Unrepresentative? ◀](#)

West Virginia Rate Regulation

- From 1993-2016, West Virginia:
 1. Capped hospital charge increases
 2. Required all hospital–insurer contracts to cover average costs
 3. Approved hospital–insurer contracts and made them public records
- Does this make West Virginia unrepresentative?
 - 1 & 2: Caps “too generous” as of Murray and Berenson (2015) and contracts easily covered costs, though may have been associated with lower list prices and more outpatient care
 - 3: disclosure unusual at time — may be more representative of where the US is going

Can Verify Larger Insurer Pays Less on Average

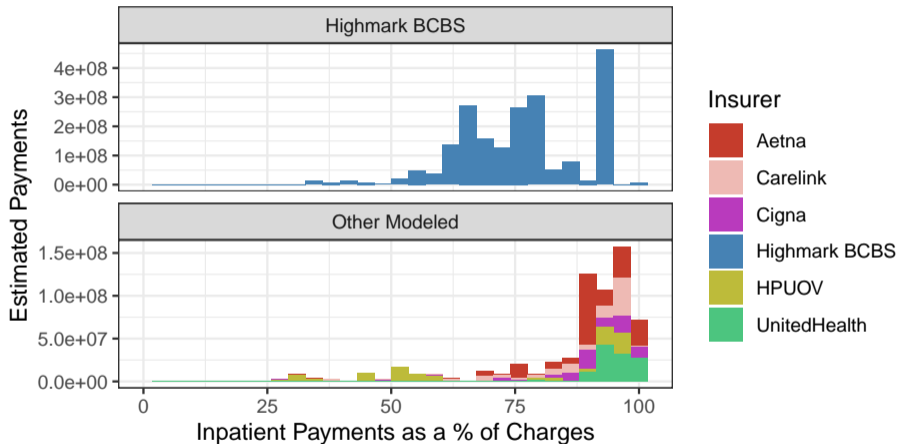


Figure: Bars are weighted by estimated payments. ◀

Network Strength

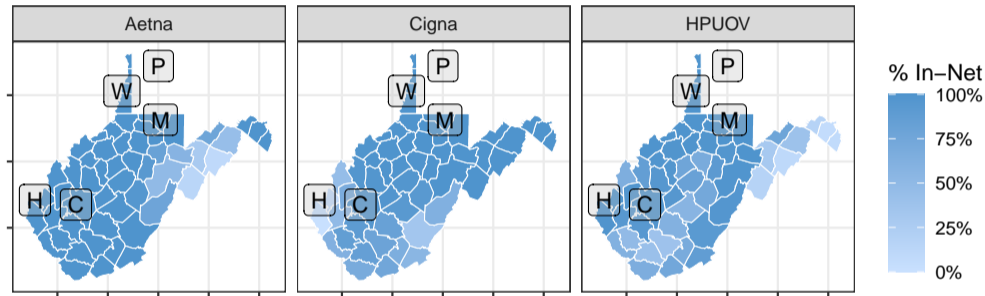


Figure: Network quality (large cities overlaid), measured as a percentage of 2016 inpatient discharges in a given insurer's 2015 network. ◀

Non-BCBS Contract Lengths (Auto-Renew)

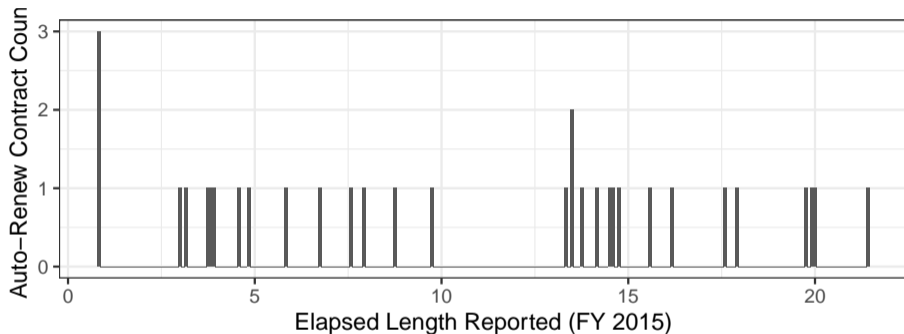


Figure: Retrospective length for non-BCBS modeled insurer auto-renew contracts (where available) as of fiscal year 2015. ◀

How I Infer Benchmarks

- Share-of-charges: same reported % of charges (up to 0.01%) in consecutive years
- Prospective (likely Medicare-based): anything else
- Possible overestimate: include per diems, any non-Medicare DRG formulas
- Possible underestimate: more charge usage than other settings (Cooper et al., 2019; Weber et al., 2019)



Benchmark Price vs. Payment Increases

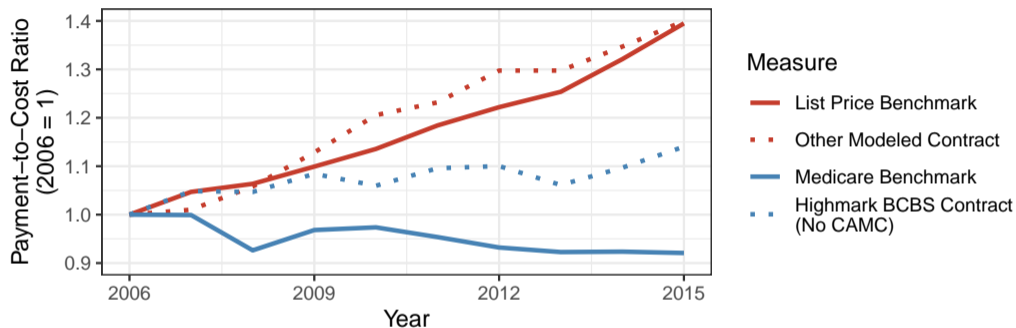


Figure: List prices went up quickly while Medicare deflated slightly relative to costs. Medium-sized insurer (list price-based) payments went up quickly while Blue Cross (Medicare-based) payments roughly tracked hospital costs. [▶ Decomposition](#) [◀](#)

Decomposition of BCBS-Other Insurer Divergence

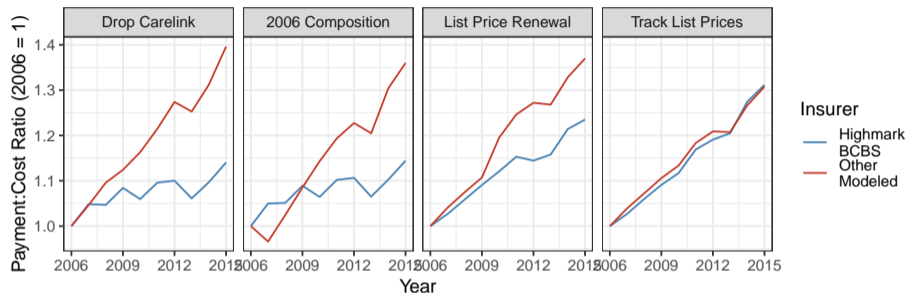
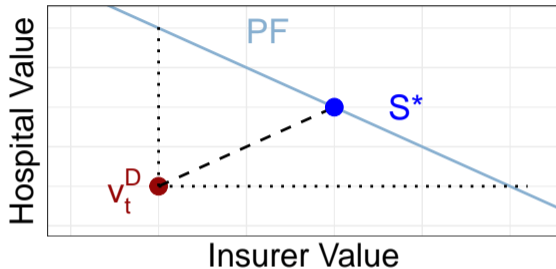


Figure: Decomposition of BCBS-other modeled insurer payment divergence by standardizing (i) insurers and (ii) insurer-hospital quantities across years, (iii) renewing prices proportionally to list prices, and (iv) renegotiating prices proportionally to list prices. The divergence between BCBS and the other insurers is largely driven by the lower renewing prices (iii) and the slower increases in starting prices (iv). ◀

Aside: What/Why is this Kalai Proportional Bargaining?

$$\frac{GFT_t^{Ins}(p_{Kalai}^*)}{GFT_t^{Hosp}(p_{Kalai}^*)} = \frac{\tau_{ij}}{1 - \tau_{ij}}$$
$$\frac{GFT_t^{Ins}(p_{Nash}^*)}{GFT_t^{Hosp}(p_{Nash}^*)} = \frac{\tau_{ij}}{1 - \tau_{ij}} \underbrace{\frac{-V_t^{Ins}(p_{Nash}^*)'}{V_t^{Hosp}(p_{Nash}^*)'}}_{=1 \text{ if TU}}$$



- Static view: “central role in the theory” (Thomson, 1994) — axioms, intuition, and data
- Generalizes Nash for **transferable utility** models like Ho and Lee (2017), but has tractability advantages for forward-looking dynamics (Dorn, 2026)

Hospital Demand Parameter Estimates

	<i>Dependent variable:</i>					
	Cancer (1)	Cardiac (2)	Digestive (3)	choice Labor (4)	Neurological (5)	Other (6)
100 Miles	-4.909** (1.949)	-8.591*** (1.597)	-7.607*** (1.821)	-29.100 (72.654)	-2.306*** (0.138)	-4.836*** (0.213)
100 Miles x Emergency	-0.409 (0.797)	-0.899*** (0.312)	-1.526*** (0.474)	4.845 (14.314)	-0.398*** (0.109)	-0.677*** (0.083)
(100 Miles) Squared	1.579** (0.644)	2.830*** (0.550)	2.295*** (0.550)	6.182 (14.056)	0.690*** (0.054)	1.518*** (0.068)
Observations	284	2,469	2,048	4,143	1,094	10,053
R ²	0.555	0.577	0.615	0.646	0.497	0.555
Log Likelihood	-286.987	-2,722.077	-2,324.572	-3,923.918	-1,297.677	-12,578.030

Note:

*p<0.1; **p<0.05; ***p<0.01

Table: Consumer valuation of distance by category in units of going from highest-value (WVU Ruby) to second highest-value (CAMC) hospital. Consumers do not like traveling, especially in non-labor emergencies. They do like flagship hospitals (omitted). ◀

Insurer Demand Estimation Details

- Estimate 2016 BCBS, Aetna, HPUOV sales based on inpatient shares
 - Ensure at least one sale per county, at least 10% of county in outside option
- Iteratively apply outer loop-inner loop strategy to find γ_k
- Outer loop for 2016 demand: take putative United and Cigna FEs given γ_k
 - Inner loop: contraction to match non-United/Cigna sales estimates with $\tilde{\delta}_{j,k,m}^{Ins} + \xi$
 - Solve for outer loop to match United & Cigna sales
 - Find new γ_k from (population) weighted least squares
- Add pre-2016 insurer FE to fit state-level sales with historical population
 - Assume Carelink had same ξ as Aetna pre-2014 acquisition

More on Bargaining Estimation

$$\sum_{t=t_0}^{t^*} \beta^{t-t_0} \text{Pay}_{ijt} - \sum_{t=t_0}^{t^*} \beta^{t-t_0} \text{Pay}_{NiN,ijt} - \text{Pay}_{NC} - \text{Pay}_{IRT} = \underbrace{\mathbb{E}_{t_0}[\omega]}_{\text{“}\omega_{ijt_0}\text{”}} = 0$$

- Moments $\mathbb{E}[\omega Z^\omega]$ and $\mathbb{E}[(\widehat{\text{MedicalLoss}} - \text{MLReport}) Z^{MLR}]$ [Details](#)
 - Z^ω hospital group & insurer dummies, Z^{MLR} insurer dummies
 - Observed flow payment Pay_{ijt} , Nash-in-Nash predicted flow payment $\text{Pay}_{NiN,ijt}$, negotiation cost payment Pay_{NC} , and “impasse repricing” transfer from forward-looking effects Pay_{IRT}
- Approximating $\text{Pay}_{IRT} \rightarrow 0$ (ij impasse $\rightarrow ij$ disagreement via others' response)

Bargaining Estimation Details

- Interpolate calendar years to bargain years via day-weighted average
- τ_{ij} : hospital system (cost) size in 2006
- Optimization in terms of bargain sets \mathcal{B} , parameters θ , and now hospital groups i :

$$\hat{\omega}_{ijt_0}(\theta) = \sum_{t=t_0}^{t^*} \beta^{t-t_0} \left(D_{ijt}^{Hosp} p_{ijt} - \widehat{PayNiN}_{,ijt}(\theta) \right) - \underbrace{\widehat{PayVC}(\theta)}_{\substack{(2\tau_{ij}-1)r_j^{Ins} \\ \text{by BCBS or not}}}$$

$$\hat{\theta} = \operatorname{argmin} \sum_j \frac{\left(\frac{1}{|\mathcal{B}_j^{Ins}|} \sum_{h,t_0 \in \mathcal{B}_j^{Ins}} \hat{\omega}_{ijt_0}(\theta) \right)^2}{\frac{\sum_{t=0}^{\operatorname{mean}(t^*-t_0)} \beta^t}{|\mathcal{B}_j^{Ins}|} \sum_{h,t_0 \in \mathcal{B}_j^{Ins}} \sum_t \beta^{t-t_0} D_{hjt}^{Hosp} p_{hjt}} + 100,000 \left(\frac{1}{6} \sum_{2011 \leq t \leq 2016} \hat{MLR}_{jt} - MLR_{jt} \right)^2$$

$$+ \sum_i \frac{\left(\frac{1}{|\mathcal{B}_i^{Hosp}|} \sum_{n,t_0 \in \mathcal{B}_i^{Hosp}} \hat{\omega}_{int_0}(\theta) \right)^2}{\frac{\sum_{t=0}^{\operatorname{mean}(t^*-t_0)} \beta^t}{|\mathcal{B}_i^{Hosp}|} \sum_{n,t_0 \in \mathcal{B}_i^{Hosp}} \sum_t \beta^{t-t_0} D_{int}^{Hosp} p_{int}}$$

Estimated Gain From Trade Shares (Myopic)

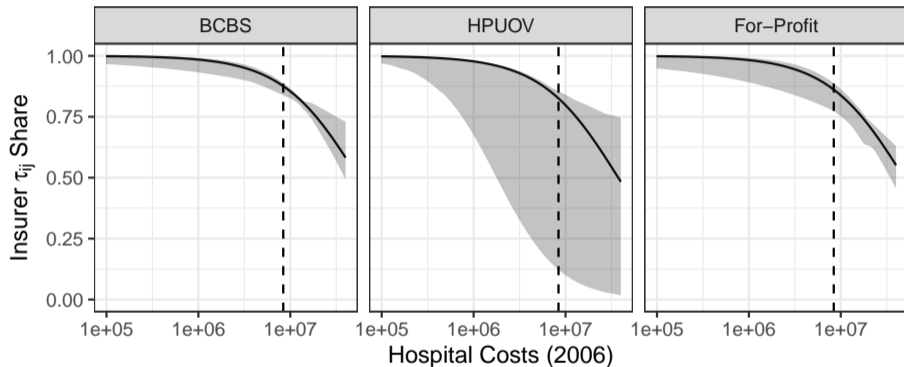


Figure: Estimated percent of gains from trade retained by the insurer under estimated myopic model. ◀

Other Bargaining Results

	Parameter (τ^{Size} Estimated)							
	η_{BCBS}	η_{HPUOV}	η_{Aetna}	$\eta_{UnitedHealth}$	η_{Cigna}	$\eta_{Carelink}$	r_{yBCBS}^M	r_{nBCBS}^M
Only-2015 (Nash/Kalai)	3657*** (45)	3404*** (85)	3658*** (116)	2008*** (29)	4627*** (32)	3139*** (39)	10000*** (2614)	9999*** (1441)
Myopic (Nash/Kalai)	4640*** (14)	4036*** (650)	3659*** (37)	3197*** (374)	4624*** (26)	3139*** (463)	10000*** (1444)	10000*** (1)
Forward-Looking ($Pay_{IRT} = 0$)	4638*** (130)	3631*** (302)	3660*** (37)	3284*** (69)	4626*** (30)	3140*** (45)	9999*** (29)	9999*** (65)
Data	3600	3356	3554	1999	4635	3114		

Note:

* $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Bargaining Model Robustness Tests

	Parameter						Parameter				
	β	τ_{BCBS}	τ_{HPUOV}	τ_{FP}	$-\tau_{Size}$		β	τ_{BCBS}	τ_{HPUOV}	τ_{FP}	$-\tau_{Size}$
Forward-Looking (Baseline)	0.899*** (0.03)	0.854*** (0.006)	0.877*** (0.026)	0.889*** (0.005)	0.989*** (0.028)	Forward-Looking (η from MLR)	0.826 (-)	0.864 (-)	0.874 (-)	0.891 (-)	0.892 (-)
Forward-Looking (No Hosp. Size)	0.714*** (0.025)	0.852*** (0.011)	0.86*** (0.01)	0.685*** (0.028)	.	Forward-Looking (Inpat. Share GFT Weight)	0.722 (-)	0.881 (-)	0.905 (-)	0.897 (-)	0.847 (-)
Forward-Looking (Mean $\sum \beta^t$ normalization)	0.925 (-)	0.854 (-)	0.876 (-)	0.89 (-)	0.991 (-)	Forward-Looking ($\beta = 0.99$)	0.99 (-)	0.854 (-)	0.875 (-)	0.881 (-)	1 (-)
Forward-Looking (Estimate Hospital Costs)	0.497 (-)	0.939 (-)	0.938 (-)	0.942 (-)	1.009 (-)	Forward-Looking (Hospital TIOLI)	0.696 (-)	0.001 (-)	0.001 (-)	0.001 (-)	.
Forward-Looking (Hospital Costs * 2)	1 (-)	1 (-)	1 (-)	1 (-)	-0.276 (-)	Forward-Looking ($\tau = 0.5$)	0.817 (-)	0.5 (-)	0.5 (-)	0.5 (-)	.
Forward-Looking (Hospital Costs * 0.9)	0.931 (-)	0.838 (-)	0.858 (-)	0.875 (-)	0.969 (-)	Forward-Looking (MCO TIOLI)	0.52 (-)	0.999 (-)	0.999 (-)	0.999 (-)	.
Forward-Looking (Hospital Costs * 1/2)	1 (-)	0.778 (-)	0.781 (-)	0.821 (-)	0.903 (-)	Myopic (Baseline)	.	0.876*** (0.012)	0.825*** (0.238)	0.861*** (0.034)	1.037*** (0.201)
Forward-Looking (Medicare Costs)	0.895 (-)	0.834 (-)	0.847 (-)	0.871 (-)	0.913 (-)	Myopic (No Hosp. Size)	.	0.863*** (0.006)	0.845*** (0.016)	0.631*** (0.028)	.

Note: *p<0.1; **p<0.05; ***p<0.01

Figure: Bargaining parameter estimates under alternative modeling assumptions. (Confidence intervals have only been implemented for some models.)

Nash-in-Kalai Moment

$$\mathbb{E}_{t_0} \left[\overbrace{\sum_{t=t_0}^{t^*} \beta^{t-t_0} \text{Pay}_{ijt}}^{\text{NPV Payment}} \right] = \mathbb{E}_{t_0} \left[\sum_{t=t_0}^{t^*} \beta^{t-t_0} \overbrace{\text{Pay}_{NiN,ijt}}^{\text{Static Nash Flow Payment}} + \overbrace{\text{Pay}_{NC} + \text{Pay}_{IRT}}^{\text{Added Multiperiod Bargaining Terms}} \right],$$

where Pay_{NC} reflects negotiation costs and Pay_{IRT} reflects the effect of spillovers on impasse profits (show zero in steady state & set to zero in empirical work). ◀

Counterfactual: Faster Medicare Rate Increases

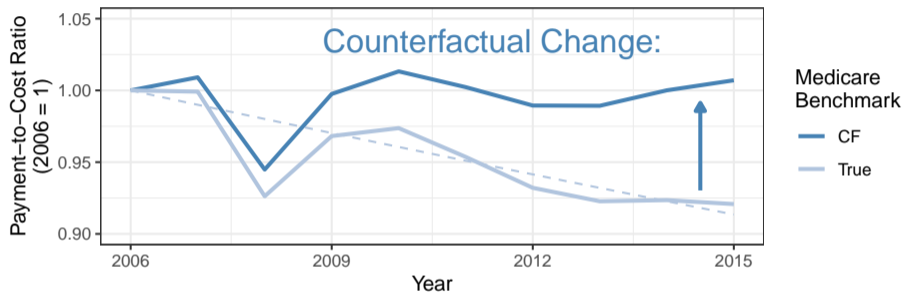


Figure: Medicare payments (light blue) decreased relative to reported costs by roughly one percentage point annually (dashed line). What if Medicare increased rates one percentage point faster each year, relative to actual rates? ◀

Counterfactual Details

- Hold renegotiation timing, benchmark choice, hospital list prices fixed
- Assume all Medicare-based payments increase one ppt faster annually
 - Content: any idiosyncratic BCBS DRG weights increase proportionally to Medicare
- Hold fixed small insurer prices (conservative)
- Main analysis holds fixed premiums (conservative)
 - Estimate downstream response from calibrated Nash-Bertrand model



A Myopic Model Would Overestimate Effects Substantially

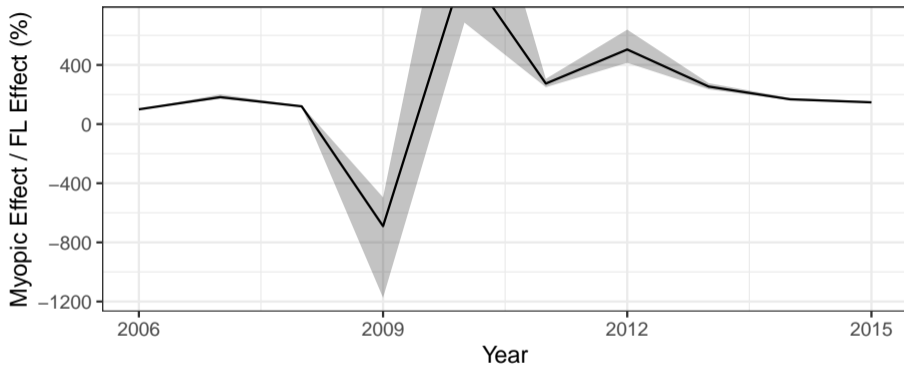



Figure: Ratio of estimated effects under myopic model to estimated effects under dynamic forward-looking model. 

Counterfactual Construction (Time Series-Adjacent)

I estimate a finite horizon model with $T = 5$ periods

$$\mathbb{E}_t \left[\sum_{t=t_0}^{t^*} \beta D_{ijt}^H \frac{p_{it}}{p_{it_0}} \frac{B_{ijt}}{B_{ijt_0}} \right] p_{ij,t_0}^R = \sum_{t=t_0}^{t^*} \mathbb{E}_{t_0} [\gamma_{ijt_0,hnt} p_{hnt}] + C_{ijt_0}$$

$$y_t \equiv \left(\mathbf{p}_{t-1}^T \quad \mathbf{p}_t^T \quad E_t[\mathbf{p}_{t+1}^T] \quad \dots \quad E_t[\mathbf{p}_{t+4}^T] \right)^T$$

$$\Gamma_{0,t} y_t = \Gamma_{1,t} y_{t-1} + C_t + \Psi_t \varepsilon_t.$$

- Changing benchmark inflation \Leftrightarrow changing Γ_1 matrices
- Estimate $\hat{\Gamma}_0$ and realized $\hat{\Gamma}_1 \Rightarrow$ realized prices in terms of $C + \varepsilon$
 - Currently set $\Psi_n = I$, recover $\widehat{C + \varepsilon}$ to match realized prices, and change Γ_1

Counterfactual: Decomposition

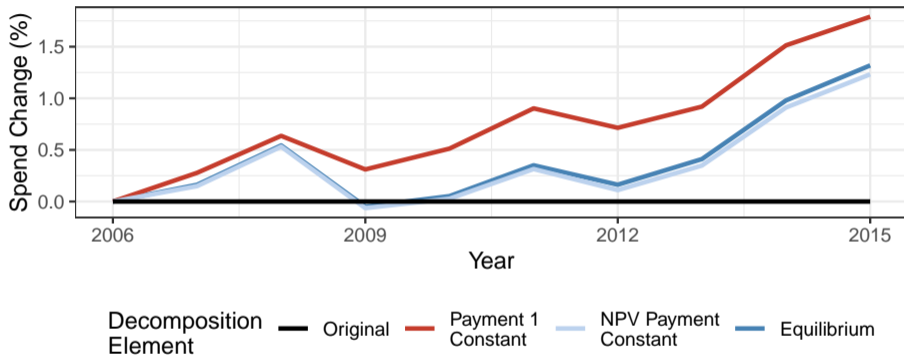


Figure: Decomposition of estimated direct effects under myopia (blue), direct effects with forward-looking firms (pink), and equilibrium effects (red).

Contracts Generally Renewed in Late Years

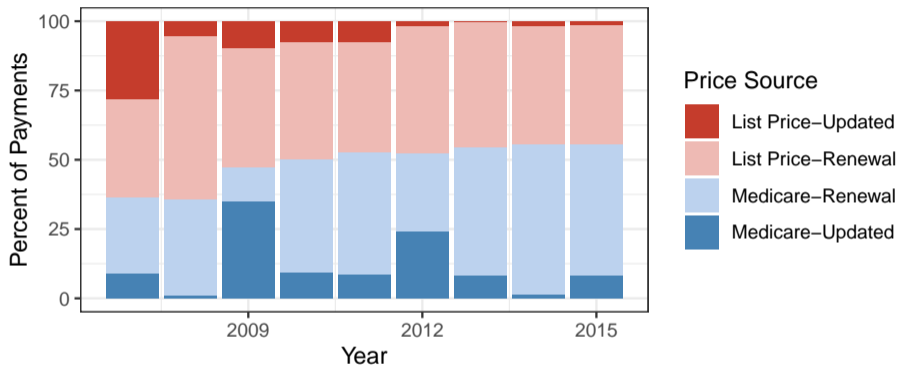


Figure: Fraction of each year's payments that reflect an inferred bargain (dark) or renewed contract (light), and which were imputed as list price-linked (teal) or other benchmark-linked (blue). Effects are large in 2015-16 because contracts generally renewed after 2013. [▶ Without imputation](#) ◀

Contracts Generally Renewed in Late Years (No Imputation)

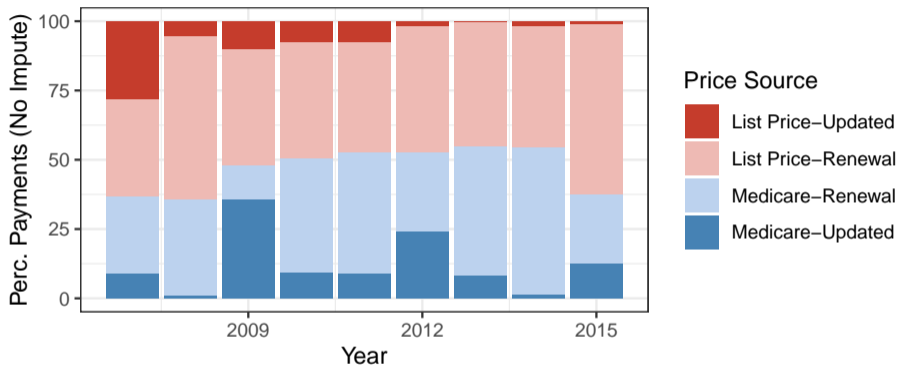


Figure: Fraction of each year's payments that reflect an inferred bargain (dark) or renewed contract (light), and which were imputed as list price-linked (teal) or other benchmark-linked (blue) without including missing reports for which I impute contracts. ◀

Counterfactual Effects by Insurer

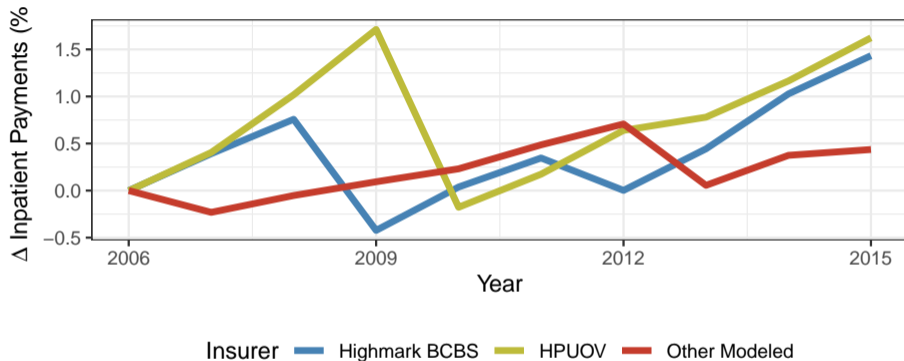


Figure: Estimated counterfactual effects on payments by insurer.

Counterfactual Effects by Hospital

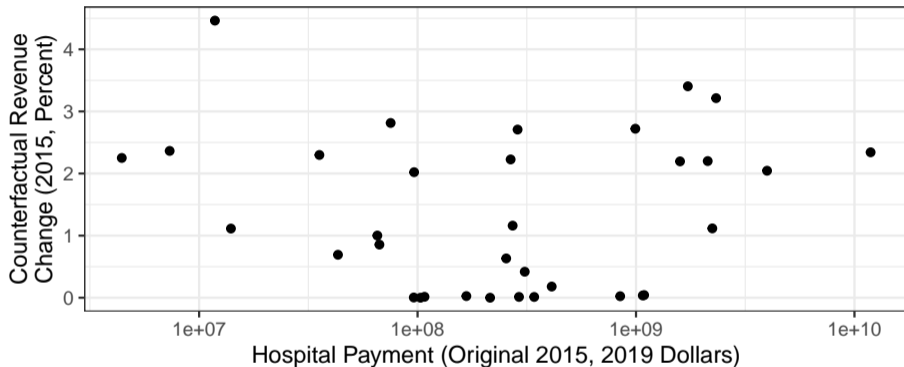


Figure: Estimated counterfactual effects on 2014 revenue by hospital. Most are generally only affected by less than 3%. 

Counterfactual: Less Discounting

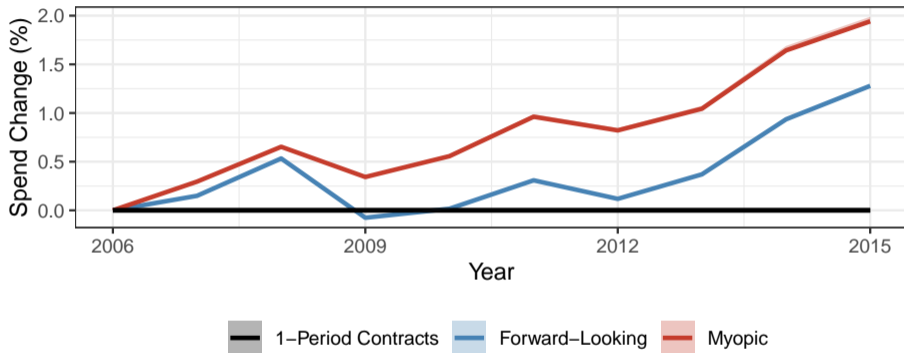


Figure: Results with discount rate β set to 0.97 rather than estimated 0.899. The forward-looking counterfactual is more forward-looking, so the estimated savings would be even smaller. ◀

Counterfactual: Downstream Premium Effects

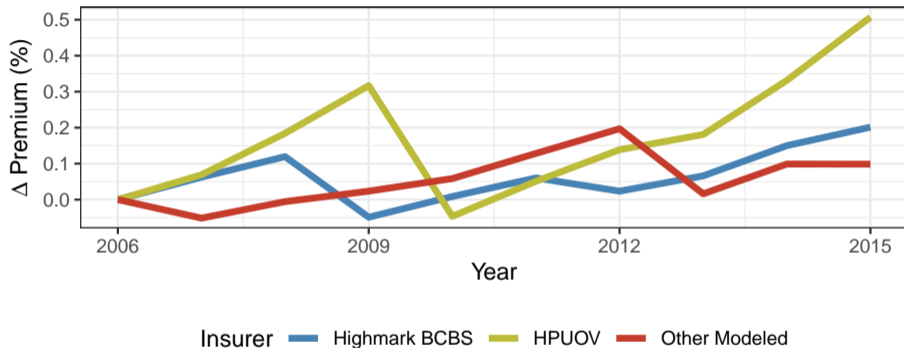


Figure: Estimated downstream effects on premiums under annual Nash-Bertrand premium competition. (Smaller percent effects because counterfactual holds outpatient constant.)

Counterfactual: Premium Change (% of Spend)

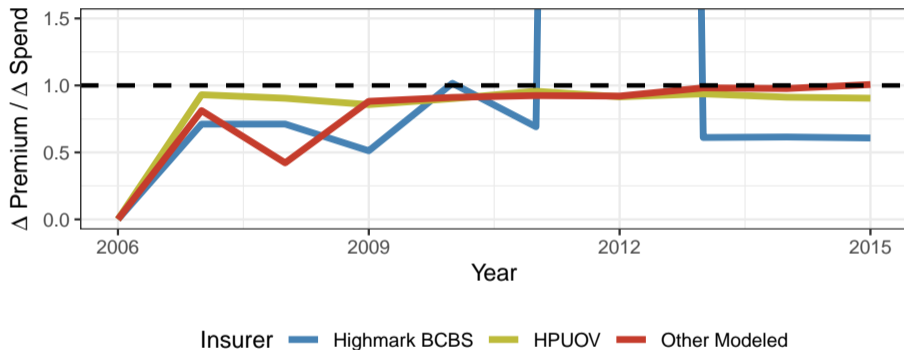


Figure: Ratio of estimated premium change to estimated marginal cost change by insurer. ◀

Counterfactual: Drop Charleston Area Medical Center

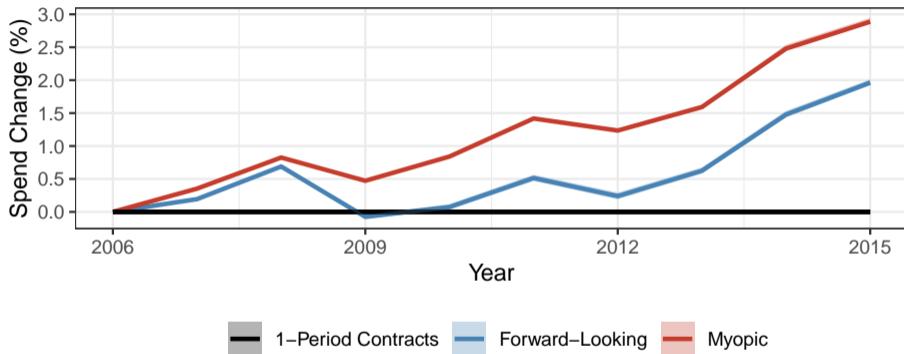


Figure: Counterfactual with Charleston Area Medical Center, a large hospital center with many low-discount high-renewal contracts, excluded. ◀

Counterfactual: List Prices Limited

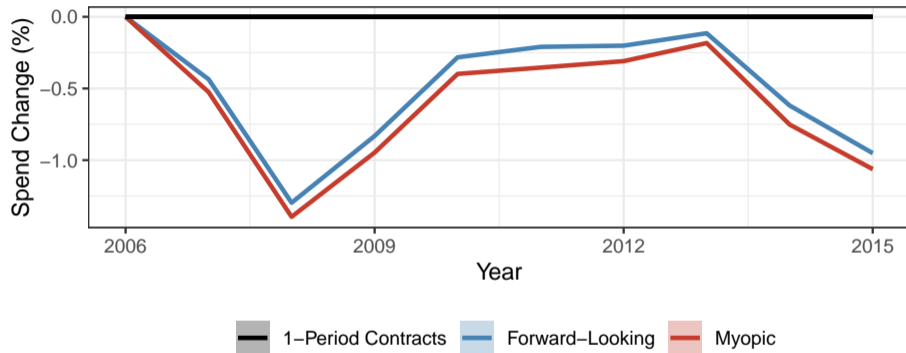


Figure: Estimated counterfactual effects on payments if list prices were capped to generally increase two ppt faster than reported costs instead of three ppt faster than reported costs. [▶ Details](#) ◀

Substantial Limitations (More in Paper)

- Miss outpatient, hold premiums & renegotiation constant
- Highly stylized insurer demand & premium-setting models (data limitations)
- Finite horizon, time definition, & end of panel biases
- Potential endogeneity of non-price bargaining like adjudication (minor)
- Effect of disagreement on consumer insurer inertia (future work)
- No effects from moral hazard, consumer cost-sharing, or benchmark choice
 - Cost-sharing small (Gowrisankaran et al., 2015), benchmark choice insurer-driven
 - Found suggestive evidence of some Medicare-driven supply effects (future work)
- Missing pre-2016 demand (data on way), investment (minimal with CON)
- Heterogeneous DRG weights, per diems, & other non-charge benchmarks
 - Shifted payments may include non-Medicare, per diems, or complex share of charges

Counterfactual List Price Construction

- Start: 102% of average state Medicare pay-to-cost change from previous year
- Where there is reliable state financial data, pull up to 50% of allowed increase towards hospital previous year Medicare increase (scaled by square root of previous year costs)
- Assume that hospitals attempt to set list prices at real list price
 - No added stockpiling effect beyond optimizing under the existing WV regulation