

The effects of centralized public procurement on prices and competition

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Public Procurement Conference – centralization and new trends

26.4.2023, Copenhagen

Introduction

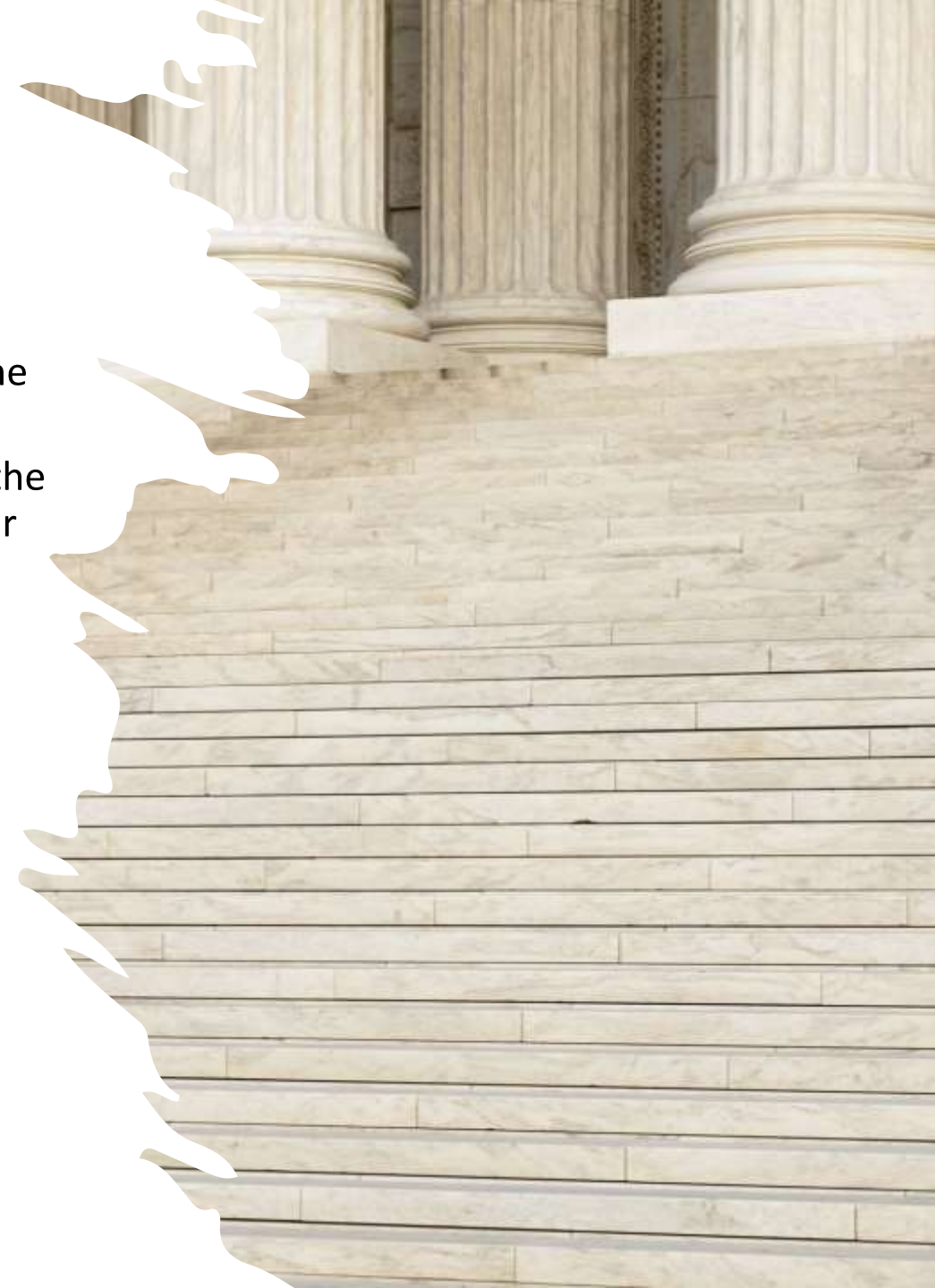
- Public procurement (PP) suffers from a **pathological lack of competition**
 - Median number of bidders is only 2 (Jääskeläinen and Tukiainen, 2019)
 - Simple auction theory and empirical evidence: **leads to high prices**
 - **Scale is huge** (13 % of GDP in OECD): even small relative efficiency gains could lead to substantial savings in public spending
- Centralized public procurement (CPP) has been proposed as one **key tool for enhancing efficiency**
 - Lower unit costs through economies of scale
 - Using better procurement skills
 - Eliminating overlapping admin costs
 - May make harder to meet the needs of individual buyers
- This chapter: CPP in the Finnish context
 - **Empirical evidence on causal effects** on prices and competition **across all industries** using comparable price measures
 - Existing evidence focuses on narrowly defined industries, mostly using Italian data. **We provide external validity on both dimensions.**

Literature

- Economics and management (see e.g. Albano & Sparro, 2010)
 - CPP is most useful for **standardized items** and services needed in **large scale** and for **local PP units** with **similar demand** and **low procurement skills**
 - Trade-offs: limited ability to tailor purchases for special needs and might result in excessively large contracts
- Game theoretic auction models: (see e.g. Dubois et al., 2021)
 - Mixed results on the price effects depending on specific assumptions
- Empirical literature (mostly in economics): (see e.g. Bandiera et al. 2009; Lotti et al. 2022)
 - Mostly based on policy-changes: introducing or mandating the use of CPP
 - **Significant negative price effects** (2-30%)
 - Not much evidence on quality effects or process costs due to lack of data

Institutional setting

- **Based on the Finnish PP law** and EU directives
 - First-price sealed bid auctions or scoring auctions, if above the tendering thresholds
 - Follows a standard procedure from preparation to choosing the contract type, publishing the tender, and choosing the winner
 - In parallel, local PP units can choose **whether to procure themselves or outsource to CPP units**
- **CPP units are defined by the law:** entities established to provide procurement services and –support for their stakeholders
 - Different from so called “in-house centralized units”
 - CPP units seem to prefer contract types that offer more flexibility (e.g. framework agreements and dynamic systems)
 - The use of CPP has been encouraged widely.
 - TED: about 15% of PP contracts procured through CPP in Finland



Data

- Data from electronic bidding platform Cloudia Oy (2013-2017 September):
 - 14,000 tenders of 204,000 procurement objects (auctions)
 - 470,000 bids and over 2 million potential bidders
 - We observe:
 - Tender characteristics (e.g. procurer name and type, tendering procedure)
 - Procurement object characteristics (e.g. CPV codes)
 - Identities of potential bidders (proxied by visiting the tender website) and actual bidders
 - Submitted bids (in euros), and the winning bid(s)
- We identify CPP units manually by their names
 - 15 units from a total of about 300
 - We cannot identify "in-house" centralized units



Defining prices

- **We aim for comparability across products and industries**
 - The observed industry classification (CPV code) is not precise enough for unit prices in most cases
 - Need to define relative prices
- **Win margin** = $(\text{second lowest bid} - \text{lowest bid}) / \text{second lowest bid}$
 - Measures the intensity of competition
 - Comparable across products and industries
 - Can be calculated for auctions with at least two bids (majority of the data)
 - Does not correlate perfectly with winning bids
- **Relative price difference** = $(\text{engineer estimate} - \text{realized price}) / \text{engineer estimate}$
 - More accurate price measure than win margin
 - BUT, can be calculated only at tender level, and for a small subset of data with both reported engineer estimates and quantities.

Figure 1: Share of CPP by industry and region

- Lack of regional variation due to multiple reasons
- Data still rich enough!

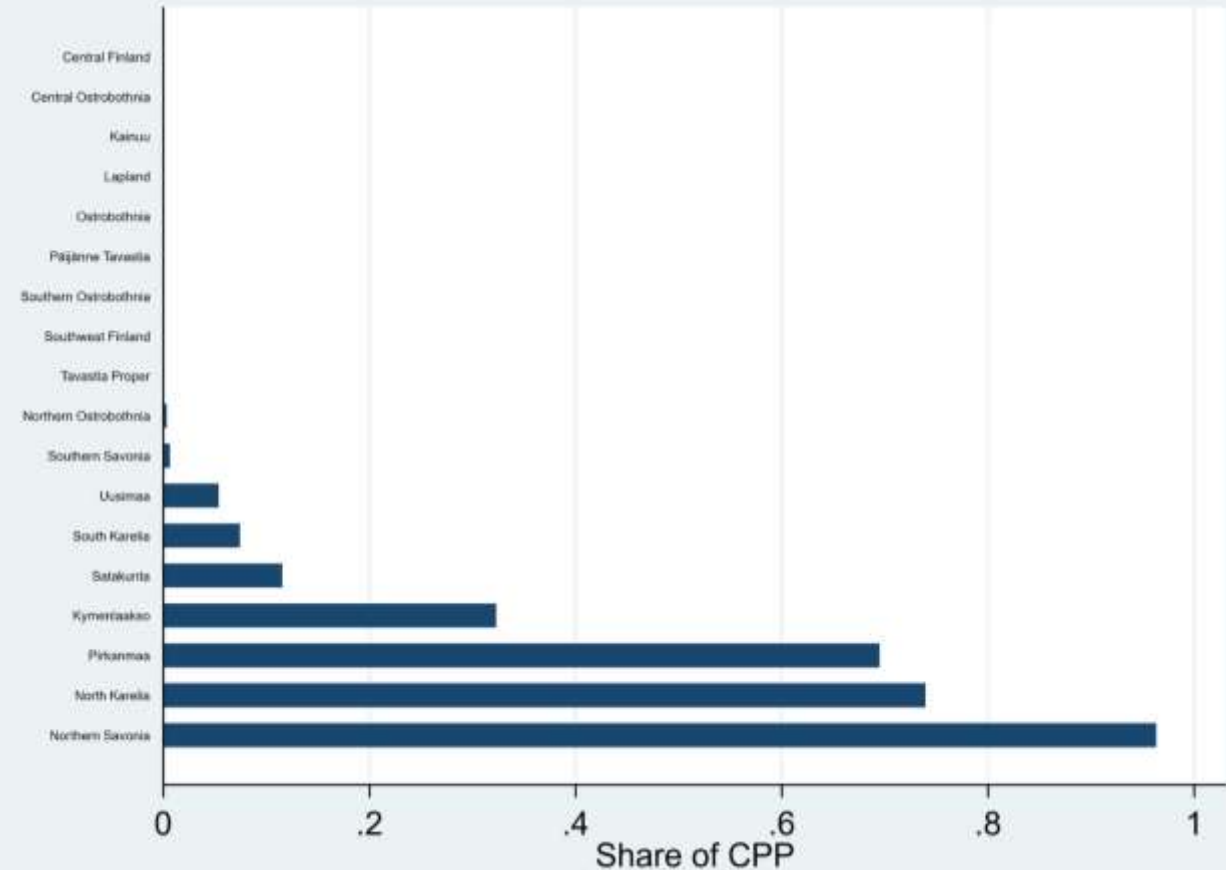
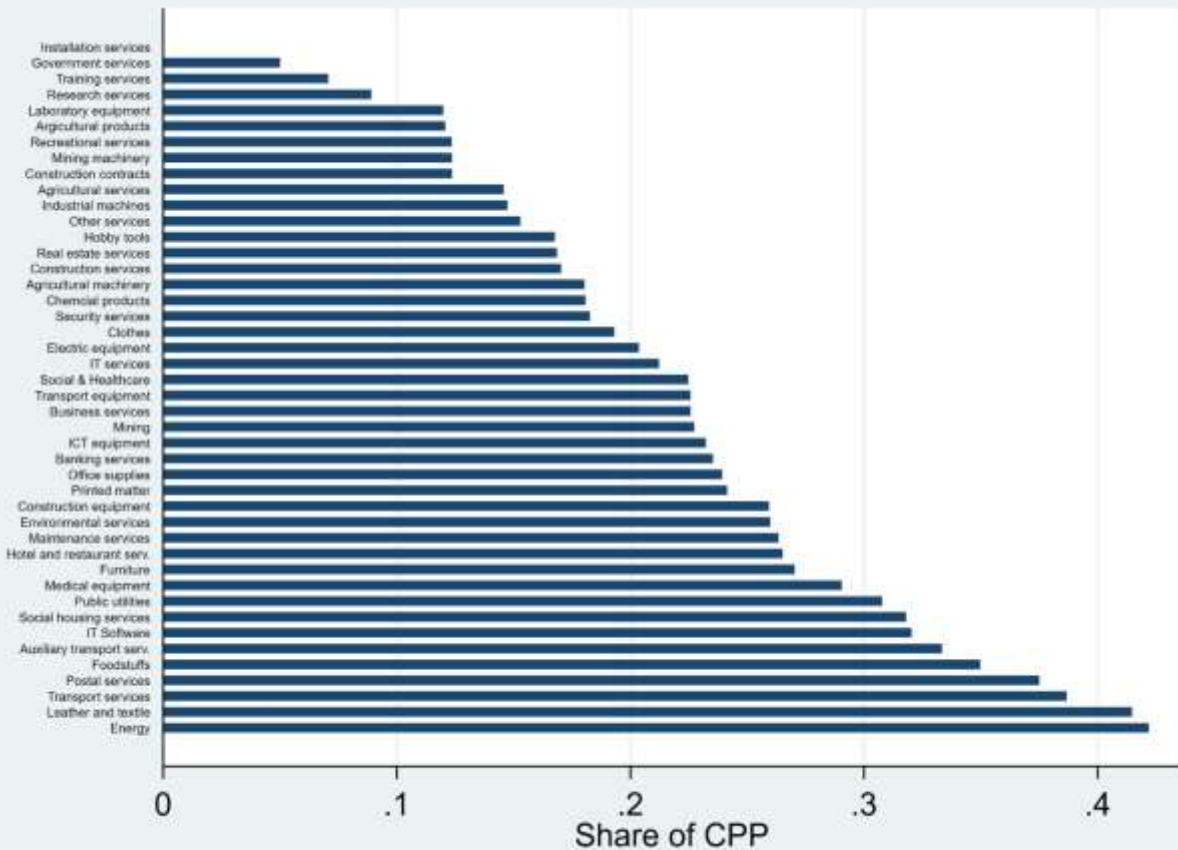
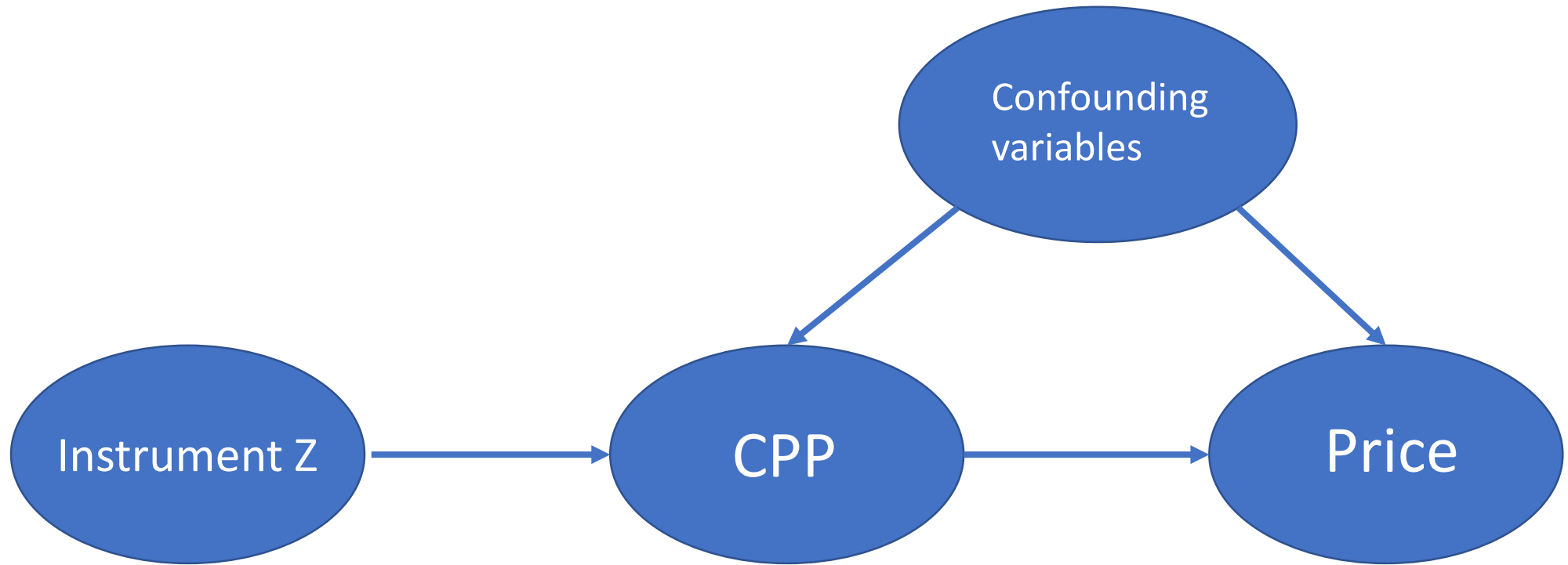


Table 2: Means of relevant variables by CPP status

Panel A: Full data						
	Tender	Tender	Tender	Auction	Auction	Auction
	All	CPP=0	CPP=1	All	CPP=0	CPP=1
CPP	0.21	0.00	1.00	0.26	0.00	1.00
Win margin	.	.	.	0.24	0.24	0.23
Rel. price difference	0.37	0.35	0.44	.	.	.
n	2.47	2.38	2.84	2.13	2.18	1.99
N	5.99	5.82	6.63	7.53	7.34	8.08
n/N	0.44	0.43	0.44	0.32	0.33	0.28
No bids	0.34	0.35	0.28	0.37	0.37	0.37
Scoring auction	0.50	0.50	0.52	0.48	0.45	0.54
Partial bidding allowed	0.25	0.23	0.32	0.51	0.49	0.58
Procurement objects	14.66	13.78	17.87	.	.	.

Method: **instrumental variable** approach



Defining an instrument

- Idea: come up with an instrument Z that affects the choice of CPP, but not the outcome P. This helps to uncover the causal relation of interest (CPP to P)
- Our instrument: **Region-level centralizing norms**
 - Rationale: there probably exists similar procedures across different tenders within a location despite contracting for different types of goods or services.
 - Calculated for each CPV-region pair leaving out information on own industry
 - Assumption 1: The norms affect the choice of using CPP, but P only through CPP.
 - Assumption 2: The instrument is correlated with CPP conditional on other covariates. This is testable.



Results – first stage

Panel B. Dependent variable: CPP (tender level)

	(1)	(2)	(3)
	OLS	OLS	OLS
Instrument Z	1.018*** (0.154)	0.969*** (0.150)	0.975*** (0.144)
Scoring auction			0.005 (0.012)
Partial bids allowed			-0.001 (0.023)
Engineer estimate disclosed			-0.087** (0.041)
Number of OBS	13317	12255	10305
Mean(CPP)	0.223	0.230	0.241
R-sq	0.604	0.653	0.702
F-stat	43.423	41.651	45.598
Industry FE	No	Yes	Yes
Year FE	No	Yes	Yes
Regional controls	No	Yes	Yes

Results – Win margin

Dependent variable: Win margin (auction level)						
	(1)	(2)	(3)	(4)	(5)	(6)
	OLS	OLS	OLS	IV	IV	IV
CPP	-0.011 (0.011)	-0.010 (0.011)	-0.005 (0.013)	-0.025** (0.012)	-0.034* (0.020)	-0.030 (0.021)
Scoring auction			0.012 (0.010)			0.013 (0.010)
Partial bids allowed			-0.035 (0.024)			-0.036 (0.025)
Engineer estimate disclosed			-0.000 (0.013)			-0.002 (0.012)
Number of OBS	99932	91102	77315	99932	91102	77315
Mean(Win margin)	0.236	0.234	0.236	0.236	0.234	0.236
R-sq	0.000	0.078	0.083	.	0.077	0.083
Industry FE	No	Yes	Yes	No	Yes	Yes
Year FE	No	Yes	Yes	No	Yes	Yes
Regional controls	No	Yes	Yes	No	Yes	Yes

Results – Relative price difference

Dependent variable: Relative price difference (tender level)						
	(1)	(2)	(3)	(4)	(5)	(6)
	OLS	OLS	OLS	IV	IV	IV
CPP	0.081*** (0.029)	0.057* (0.031)	0.059** (0.027)	0.061* (0.033)	0.049 (0.033)	0.037 (0.036)
Scoring auction			-0.019 (0.017)			-0.019 (0.017)
Partial bids allowed			-0.051 (0.060)			-0.052 (0.059)
Number of OBS	1354	1217	1112	1354	1217	1112
Mean(Rel. Price difference)	0.374	0.374	0.365	0.374	0.374	0.365
R-sq	0.017	0.153	0.210	0.016	0.152	0.210
Industry FE	No	Yes	Yes	No	Yes	Yes
Year FE	No	Yes	Yes	No	Yes	Yes
Regional controls	No	Yes	Yes	No	Yes	Yes

Results – Number of actual bidders

Panel B: Dependent variable: Number of actual bidders (tender level)

	(1)	(2)	(3)	(4)	(5)	(6)
	OLS	OLS	OLS	IV	IV	IV
CPP	0.440*	0.332	0.282	0.425	0.315	0.268
	(0.232)	(0.211)	(0.222)	(0.291)	(0.304)	(0.312)
Scoring auction			-0.247**			-0.247**
			(0.100)			(0.099)
Partial bids allowed			0.561**			0.560**
			(0.245)			(0.243)
Engineer estimate disclosed			-0.062			-0.065
			(0.126)			(0.133)
Number of OBS	13317	12255	10305	13317	12255	10305
Mean(n)	2.499	2.531	2.519	2.499	2.531	2.519
R-sq	0.005	0.049	0.060	0.005	0.049	0.060
Industry FE	No	Yes	Yes	No	Yes	Yes
Year FE	No	Yes	Yes	No	Yes	Yes
Regional controls	No	Yes	Yes	No	Yes	Yes

Results – Number of potential bidders

Panel B: Dependent variable: Number of potential bidders (tender level)

	(1)	(2)	(3)	(4)	(5)	(6)
	OLS	OLS	OLS	IV	IV	IV
CPP	0.710** (0.312)	0.968*** (0.251)	0.827*** (0.280)	0.518 (0.411)	1.027*** (0.386)	0.865** (0.401)
Scoring auction			0.315* (0.165)			0.316* (0.164)
Partial bids allowed			1.286*** (0.395)			1.287*** (0.392)
Engineer estimate disclosed			-1.132*** (0.186)			-1.127*** (0.191)
Number of OBS	13317	12255	10305	13317	12255	10305
Mean(N)	6.081	6.104	5.829	6.081	6.104	5.829
R-sq	0.004	0.102	0.132	0.004	0.102	0.132
Industry FE	No	Yes	Yes	No	Yes	Yes
Year FE	No	Yes	Yes	No	Yes	Yes
Regional controls	No	Yes	Yes	No	Yes	Yes

Results – Entry rate

Panel B: Dependent variable: Entry rate (tender level)

	(1)	(2)	(3)	(4)	(5)	(6)
	OLS	OLS	OLS	IV	IV	IV
CPP	0.010 (0.032)	-0.026 (0.026)	-0.016 (0.025)	0.009 (0.038)	-0.054 (0.039)	-0.043 (0.038)
Scoring auction			-0.056*** (0.012)			-0.056*** (0.011)
Partial bids allowed			-0.017 (0.036)			-0.018 (0.036)
Engineer estimate disclosed			0.090*** (0.016)			0.086*** (0.017)
Number of OBS	12427	11413	9623	12427	11413	9623
Mean(Entry rate)	0.435	0.439	0.455	0.435	0.439	0.455
R-sq	0.000	0.057	0.068	0.000	0.057	0.067
Industry FE	No	Yes	Yes	No	Yes	Yes
Year FE	No	Yes	Yes	No	Yes	Yes
Regional controls	No	Yes	Yes	No	Yes	Yes

Main results summary

- **A cautious interpretation of our results:** CPP seems to lead to slightly lower prices by attracting more potential bidders, from which a similar share of suppliers submit a bid, increasing slightly the number of competitors and hence the intensity of competition.
 - Any conclusions are subject to uncertainty due to lack of statistical power
- In line with prior literature, but magnitudes significantly lower. Possible reasons:
 - Existing studies from countries that are culturally/institutionally different from Finland
 - Indirect effects of CPP
 - CPP might work much better in different industries
- IV not suitable for single industry, but we have correlational evidence on cleaning industry
 - Price correlations are large

Conclusion

- We study CPP's effects on prices and competition empirically with rich Finnish PP data
 - IV approach to identify causal effects
 - Normalized price measures
 - Across different industries
- Results:
 - CPP induces only modest negative price effects overall.
 - CPP does seem to enhance competition.
 - Despite using large data, there is considerable statistical uncertainty
 - Given the scale of PP at large, even small efficiency gains could have significant implications in absolute terms.
- Although qualitatively aligned with previous empirical evidence, our estimates are much lower in magnitude.
 - We discuss multiple possible explanations
- Discussed arguments do not tackle the fact that PP is working seemingly ineffectively in Finland as a whole, which calls for a change in PP design and policy.
- **Given the discussed and demonstrated strengths and weaknesses of CPP, centralization seems to have only a partial role in this process.**

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