

# Changing Market Structure and Evolving Ways to Compete: Evidence from Retail Gasoline

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#### Process innovation: from full service to self service



https://365v.co.kr/article/1481865330356/22000000



http://www.ekn.kr/news/article.html?no=326856



http://blog.naver.com/PostView.nhn?blogId=pmh0318&logNo=220969572024



http://www.ichannela.com/news/main/news\_detailPage.do?publishId=000000067006



# Why gasoline market?

Nice to:

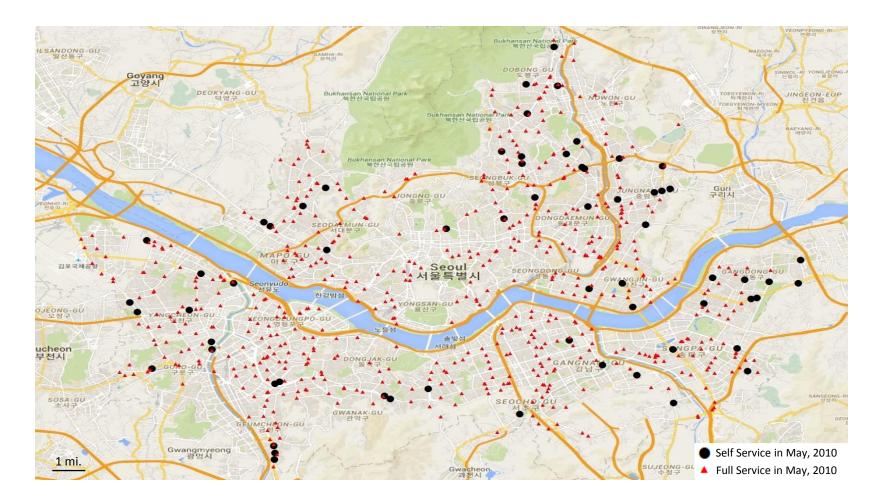
- observe the pricing behavior of sellers by product level
- examine the effect of an innovation of self-service technology on price



http://bundling.tistory.com/84



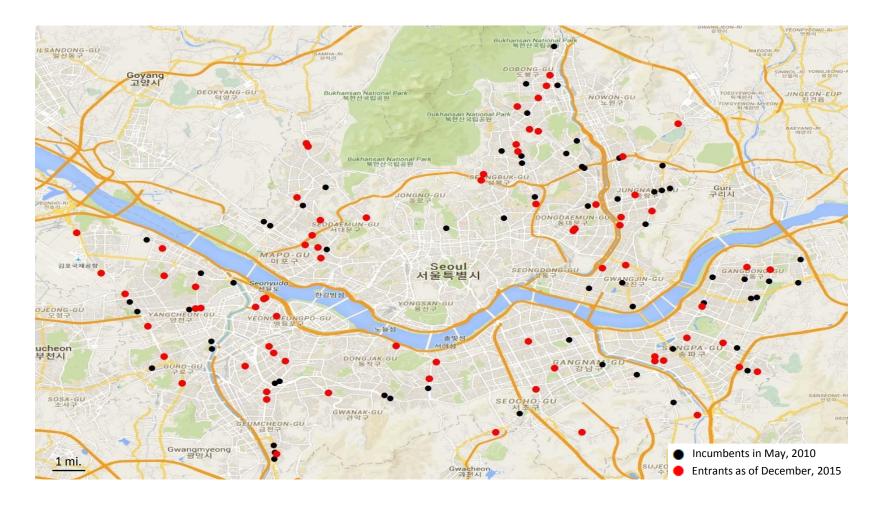
#### Research motivation: Market transition



Location of Gasoline Stations in Seoul



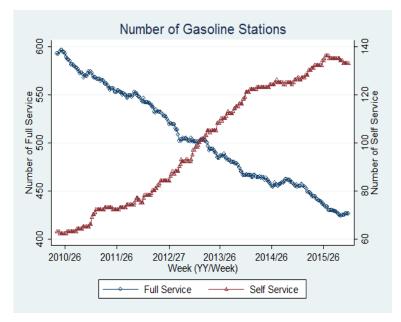
#### Research motivation: Market transition



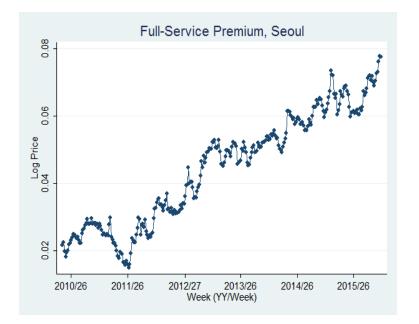
Expansion of Self-Service Stations in Seoul



#### Stylized fact: the evolution of full-service premium



The Change in the Number of Stations

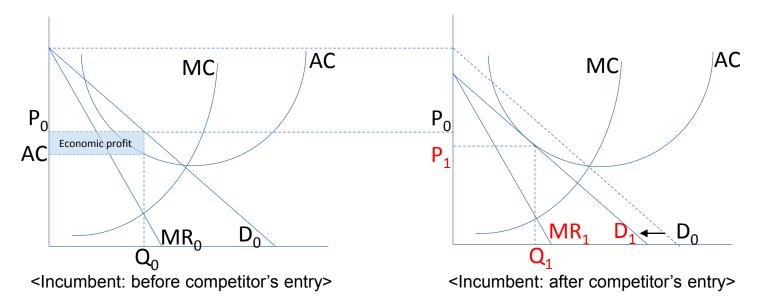


The Increasing Full-Service Premium



#### **Research question**

- Why does the price gap increase during the transition?
  - OBVIOUS if the F-S vs. S-S markets are sufficiently segmented
- Insight from general models of monopolistic competition
  - Positive profit gives another competitor an incentive to enter (i.e., self-serve market)
  - The entry shifts incumbent's residual demand to the left  $\rightarrow$  price falls





# **Preview of findings**

- 1) Competition story alone isn't enough to explain the gap
  - Not OBVIOUS: the markets are not segmented during the transition
  - based on difference-in-difference estimation
- 2) Confirmation of competitive effects on prices
  - (direct effect) self-service sellers offer lower price
  - (indirect effect) self-service sellers fall their nearby competitors
  - No significant effect of full-service sellers on prices
- 3) Evidence on competition through product differentiation
  - a higher premium to consumers who buy a bundle of gas and services
  - descriptive evidence on product differentiation and supplementary analyses



### Data

- Overview
  - Daily station-level data from May 2010 to December 2015, from OPINET
    - ✓ Wednesday only (but the loss of the information is minimal)
  - Transactions data, automatically uploaded at OPINET
  - Station characteristics: brand, service, multiproduct, location
  - Information on stations' promotion & price, collected on May 2017
- Notes
  - Inferred information on entry/exit from the price-reporting regulation
  - Two measures of competitive conditions: numbers & distances

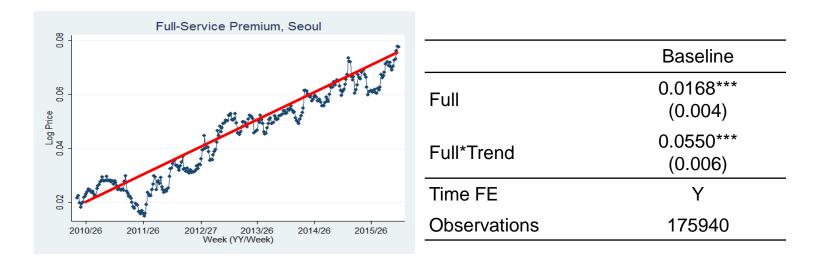
	Mean	SD	Min	Max	Mean	SD	Min	Max
Number of	Full-S	ervice	e Com	petitors	Self-S	ervice	e Com	petitors
Within 0.5 mile $(\#)$	1.98	1.54	0.00	10.0	0.41	0.67	0.00	4.0
Within 1.0 mile $(\#)$	7.15	3.44	0.00	21.0	1.33	1.32	0.00	7.0
Distance to	Full-S	ervice	e Com	petitors	Self-S	ervice	e Comj	petitors
1st nearest (mi)	0.32	0.22	0.01	2.70	0.89	0.63	0.02	3.50
2nd nearest (mi)	0.50	0.24	0.03	2.71	1.31	0.70	0.12	4.12



# **Empirical strategy**

#### is to find economic variables making $\phi$ insignificant

- Static concept explanation when assuming segmented market
  - Increase in # S-S stations  $\rightarrow$  High competition between S-S
  - Decrease in # F-S stations  $\rightarrow$  Low competition between F-S
  - → It predicts an increasing price gap between F-S and S-S
- Identify the increasing price gap b/w F-S vs. S-S
  - $\ln P_{it} = \theta Full_{it} + \phi (Full_{it} * Trend_t) + \mu_t + \varepsilon_{it}$  where Trend  $\in (0,1]$





# **Empirical strategy**

- Hypothesis #1: station characteristics with systematical correlation
  - $\ln P_{it} = \zeta X_{it} + \delta_i + \theta Full_{it} + \phi (Full_{it} * Trend_t) + \mu_t + \varepsilon_{it}$ 
    - X: covariates including brands, brand share, multiproduct.
- Hypothesis #2: competition story (two competition measures)
  - $\ln P_{it} = \beta_1 \operatorname{Num}_{it}^{SS} + \beta_2 \left( \operatorname{Num}_{it}^{SS} * \operatorname{Full}_{it} \right) + \beta_3 \operatorname{Num}_{it}^{FS} + \beta_4 \left( \operatorname{Num}_{it}^{FS} * \operatorname{Full}_{it} \right)$  $+ \zeta X_{it} + \delta_i + \theta \operatorname{Full}_{it} + \varphi (\operatorname{Full}_{it} * \operatorname{Trend}_t) + \mu_t + \varepsilon_{it}$
  - $\ln P_{it} = \beta_1 \ln Dist_{it}^{SS} + \beta_2 (\ln Dist_{it}^{SS} * Full_{it}) + \beta_3 \ln Dist_{it}^{FS} + \beta_4 (\ln Dist_{it}^{FS} * Full)$ +  $\zeta X_{it} + \delta_i + \theta Full_{it} + \phi (Full_{it} * Trend_t) + \mu_t + \varepsilon_{it}$

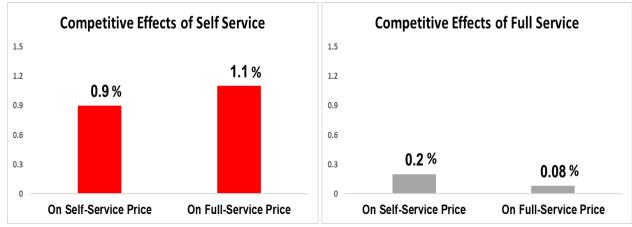


### **Result #1:** competition story doesn't explain well

	Baseline	Controls Included	Nums included	Distances Included
Full	0.0168*** (0.004)	0.0196*** (0.006)	0.0231** (0.010)	0.0168* <u>(0.010)</u>
Full*Trend	0.0550*** (0.006)	0.0477*** (0.006)	0.0486*** (0.006)	0.0495*** (0.006)
Station FE	Ν	Y	Y	Y
Controls	Ν	Y	Y	Y
Observations	175940	175940	175940	175940

Robust standard errors in parentheses, clustered by station

• Competitive effects: when the distance to nearby competitors doubles

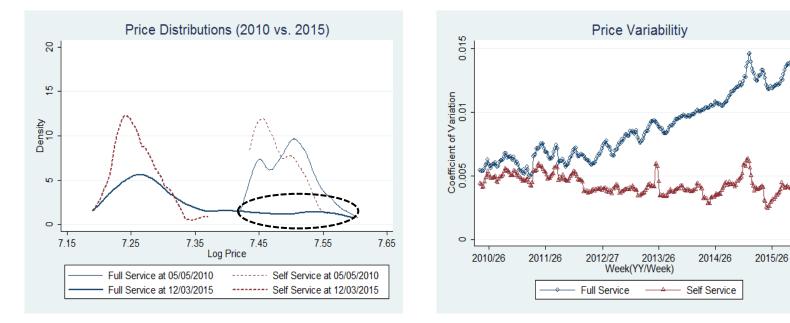


Red: significance at the 1% level & Grey: insignificance at the 10% level



# Stylized fact: the evolution of full-service premium

• Take a <u>closer</u> look at the increasing full-service premium



Price distributions on 2010 vs. 2015

Time Series of Coefficient of Variation



- In which theory a seller charges a higher premium?
- Product differentiation and its softening effect on competition
  - Mazzeo (2002, RESTAT), motel industry
  - Basker & Noel (2009, JEMS) and Matsa (2011, QJE), supermarket
- Search friction and its softening effect on competition
  - Sorensen (2000, JPE) and Ching (2010, IJIO), pharmaceutical market
  - Lewis (2011, IJIO) and Kim (2018, RIO), retail gasoline
  - Ellison and Ellison (2009, Econometrica), online shopping



(C)





Nerey Compony Blog



(b)



• Provide "free" bundled offers (collected on May 17, 2017)

	# Full Service	# Self Service
Reward points	29	4
Carwash	9	3
Carwash & Coffee	1	0
Carwash & Coffee & Washer fluid	1	0
Coffee or Tea	3	1
Coffee & Washer fluid	1	0
Coffee & Facial tissue	2	0
A bottle of water	3	0
Facial tissue	1	1
A bottle of water or Facial tissue	1	1
Car Inspection	2	0
Service for Diplomatic vehicle	1	0
Coffee, Soda, Noodle, Copy/Fax, Lounge	0	1
Total #: 539 stations	<b>55</b> /392 (14%)	<b>11</b> /147 (7%)

		Full Service		$\mathbf{Self}$	Service
	May 17, 2017	# Stations	Avg. Log Price	e # Stations	Avg. Log Price
	Bundled	55	7.444	11	7.321
	Not Bundled	337	7.360	136	7.298
bundling premium 🚽	Difference		0.085		0.023
0.	P-value		(0.000)		(0.073)
	All	392	7.372	147	7.300
	Difference			$0.072 \leftarrow full-ser$	vice premium
	P-value			0.000)	•



- Supplement descriptive evidence using my price panel
- Identify stations in my sample that provide
  - 1. bundled products in 2017
  - 2. not bundled products in 2017

	Full	Full Service		Self Service		
May 05, 2010	# Stations	Avg. Log Pric	e # Stat	tions Avg. L	og Price	
Bundled in 2017	48	7.514	2	7.	503	
Not bundled in 2017	290	7.499	50	) 7.	475	
Difference	little	→ 0.015		0.	028	
P-value		(0.011)		(0.	268)	
All	338	7.501	52	2 7.	476	
Difference			0.025			
P-value			(0.000) <b>fu</b>	<u>ll-service p</u>	<u>remium</u>	

	Full	Full Service		Self	Service
Dec 16, 2015	# Stations	Avg. Log Price	e #	Stations	Avg. Log Price
Bundled in 2017	55	7.407		7	7.282
Not bundled in 2017	332	7.326		115	7.258
Difference	significant	0.081			0.023
P-value	- 0	(0.000)			(0.156)
All	387	7.338	_	122	7.259
Difference			0.078	full cor	vice premium
P-value		(	(0.000)	iuii-sei	vice premium



- Supplement the descriptive evidence, using price data
- Examine the stability of price rankings
  - Search models
    - $\checkmark$  Uninformed consumers and price dispersion for one homogenous good
  - Insight from models of search
    - ✓ Customers search for a low price, so sellers' relative prices change from one time to the next
  - Hypothesis
    - Sellers' relative prices should go up and down <u>if a product of sellers is homogenous</u> (holding other station characteristics constant)
- Create price rankings for each week and calculate transition probabilities



					T+1					
		1	2	3	4	5	6	7	8	Total
Lowest 12.5%	1	77.35	20.03	2.14	0.35	0.09	0.04	0.00	0.00	100.0
	2	18.32	59.12	20.58	1.69	0.21	0.07	0.00	0.02	100.0
	3	2.73	18.19	59.71	18.25	0.97	0.11	0.02	0.01	100.0
Т	4	0.67	2.38	15.30	66.23	14.92	0.45	0.04	0.01	100.0
	5	0.28	0.43	1.74	12.83	73.93	10.56	0.20	0.03	100.0
	6	0.15	0.13	0.20	0.68	9.37	81.04	8.37	0.07	100.0
	7	0.04	0.06	0.08	0.17	0.37	7.43	86.53	5.35	100.0
Highest 12.5 %	8	0.06	0.05	0.05	0.05	0.07	0.28	4.87	94.58	100.0
Total		12.37	12.56	12.48	12.56	12.46	12.45	12.52	12.60	100.0

#### Price Octile Transition Matrix, from one week to the next



					T+1					
		1	2	3	4	5	6	7	8	Total
Lowest 12.5%	1	77.35	20.03	2.14	0.35	0.09	0.04	0.00	0.00	100.0
	2	18.32	59.12	20.58	1.69	0.21	0.07	0.00	0.02	100.0
	3	2.73	18.19	59.71	18.25	0.97	0.11	0.02	0.01	100.0
т	4	0.67	2.38	15.30	66.23	14.92	0.45	0.04	0.01	100.0
	5	0.28	0.43	1.74	12.83	73.93	10.56	0.20	0.03	100.0
	6	0.15	0.13	0.20	0.68	9.37	81.04	8.37	0.07	100.0
	7	0.04	0.06	0.08	0.17	0.37	7.43	86.53	5.35	100.0
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Total		12.37	12.56	12.48	12.56	12.46	12.45	12.52	12.60	100.0

#### Price Octile Transition Matrix, from one week to the next

✓ Relative prices are more stables for higher-priced stations



• Ranking stability test:

$$\begin{split} \text{Stable}_{it} &= \beta \text{Octile}_{it-1} + \theta \text{Full}_{it} + \gamma \mathbf{X}_{it} + \delta_i + \mu_t + \epsilon_{it} \\ \text{where stable} &= \begin{cases} 1 & \text{if Octile}_{it-1} = \text{Octile}_{it}; \\ 0 & \text{if Octile}_{it-1} \neq \text{Octile}_{it}. \end{cases} \end{split}$$

	Baseline	By Service
Octilo	0.0296***	0.0092*
Octile	(0.002)	(0.005)
Octilo*Eull		0.0251***
Octile*Full		(0.006)
Full	0.0433***	-0.0390***
Full	(0.016)	(0.023)
Num <sup>ss</sup>	-0.0000	0.0003
Nullies	(0.004)	(0.004)
Num <sup>FS</sup>	0.0095***	0.0104***
	(0.0023)	(0.0023)
Station and Time FE	Yes	Yes
Observations	174648	174648
% predicted outside [0, 1]	2%	3%

# Conclusion

- By service level, different strategic choices are chosen:
  - F-S stations increasingly differentiate their product to compete for less-pricesensitive consumers
  - S-S stations decreases their local competitors' price and compete for pricesensitive consumers
- We do not know what this market will look like in the future

"기름을 넣는곳 → 기름도(!) 넣는곳"

택배 받고 자동결제까지... 융합 서비스 다각화 '스마트스테이션'

수소, 전기, 휘발유, 경유, LPG 연료를 한 곳에서 채울 수 있는 '복합에너지스테이션'

#### "There's a shift in people buying gas based on the quality of the sandwich as opposed to getting a sandwich based on the price of gas,"

- Vice President for NACS in Bloomberg interview (Aug 18, 2017) \*NACS: National Association of Convenience Stores in US 감사합니다



### Appendix: summary statistics

Variable	Description	Mean	SD	Min	Max
Р	Price of gasoline (unit: KRW/liter)	1904.7	206.7	1317	2490
$\ln P$	Log price of gasoline	7.54	0.11	7.18	7.82
Full	Station offering full-serve gasoline	0.83	0.37	0	1
SK	Station brand: SK Energy	0.36	0.48	0	1
$\operatorname{GS}$	Station brand: GS Caltex	0.25	0.43	0	1
SO	Station brand: S-Oil	0.11	0.31	0	1
HD	Station brand: Hyundai Oilbank	0.12	0.33	0	1
AD	Station brand: Alddle	0.02	0.13	0	1
Unbranded	Station brand: Unbranded	0.03	0.17	0	1
Brand Share	Share of same-brand stations within 1.5 miles	0.27	0.16	0	1
Multi	Station selling regular and premium gasoline	0.33	0.47	0	1
Store <sup>a</sup>	Station having a convenience store	0.09	0.28	0	1
$Carwash^{a}$	Station having an automatic carwash equipment	0.67	0.46	0	1
Repair <sup>a</sup>	Station having a auto-repair facility	0.26	0.44	0	1
$\mathrm{Income}^{\mathrm{b}}$	Household monthly income (unit: KRW million)	3.69	0.43	2.72	4.49
$Car^b$	Number of vehicles (unit: thousand)	104.0	41.8	37.2	193.65
Stations	Number of gasoline stations in Seoul	602.8	25.4	558	658

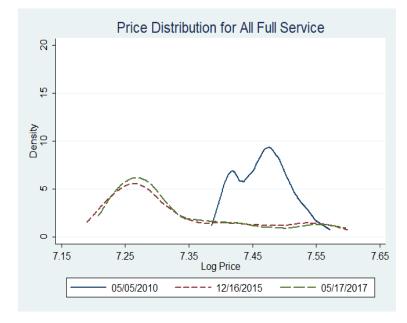
Note: Average across all stations in all time periods, except for Income and Car

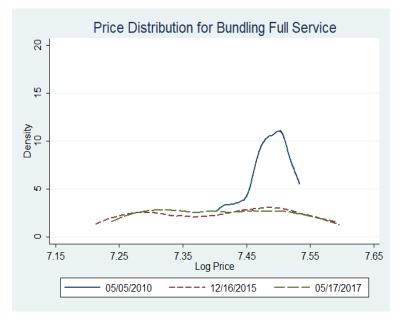
<sup>a</sup> Observations at the station level in two days; the first Wednesday of May 2010 and 2011

 $^{\rm b}$  Observations at the district level in two months; May 2010 and May 2011



### Appendix: price distribution comparison







#### **Appendix:** who exits? who converts?

- Define <u>F-S stations' choices</u> based on entry & exit information
  - "permanent exit"
  - "conversion to SS"
  - "FS continuation"
- Assume one choice for each station during the sample period:
  - # of full-service stations in the market: 593 (first day)

	permanent exit	conversion to SS	FS continuation
# of instances, 2010-2015	122	65	406



#### **Appendix:** who exits? who converts?

• Station i's decision to be correlated with intensity of competition:

$$\log\left(\frac{\pi_{i,j}}{\pi_{i,j}}\right) = \alpha_j + \beta_j \text{Num}_i^{\text{SS1}} + \gamma_j \text{Num}_i^{\text{FS1}} + \zeta_j \mathbf{Z}_i + \varepsilon_j$$

where j = {"permanent exit" or "conversion to SS"} and J = "FS continuation"

- $\frac{\pi_{i,j}}{\pi_{i,j}}$ : the odds that full-service station i falls in category j as opposed to the baseline outcome
- Num<sup>SS1</sup>; Num<sup>FS1</sup>: the number of self- and full-service competitors within one mile
- Z: covariates (e.g., presence of store/carwash/repair, brand, income, car, multiproduct)
- ε: robust standard error, clustered by Numrict
- $\checkmark$  β and γ are the relative-probability ratios of one decision to "FS continuation"



#### Appendix: who exits? who converts?

• Relative probability, as opposed to "FS continuation"

	(1)	(2)
	May 05, 2010	May 04, 2011
Outcome: permanent exit		
Num <sup>SS1</sup>	1.293**	1.286***
Num	(0.145)	(0.125)
Num <sup>FS1</sup>	1.047*	1.073***
Nulli <sup>, o</sup>	(0.023)	(0.026)
Outcome: conversion to SS		
Num <sup>SS1</sup>	1.413*	1.345**
Numee	(0.267)	(0.178)
Num <sup>FS1</sup>	1.021	1.048
Num <sup>, or</sup>	(0.032)	(0.033)
Control variables	Y	Y
Observations	591	546
** n-5% · * n-10%		

\*\* p<5%; \* p<10%

#### ✓ Greater price-competition *drives* high-cost sellers *out* of a market