



Why are Prices Sticky?

A Test of Alternative Models of Price Adjustment

Christopher Douglas – University of Michigan, Flint
Ana María Herrera – Michigan State University

Index Measures Conference, May 15 2008

WHY ARE PRICES STICKY?

- Price stickiness required to drive a real money-output relationship.
 - The *form* of price stickiness is key to understanding inflation dynamics (Reis, 2006).

- “The idea that prices set by firms in concentrated industries might exhibit rigidities is an old concern of industrial-organization economists” (Slade, 1999).

- Understand pricing patterns on the microeconomic level to understand how best to model them on the macroeconomic level.



A ROADMAP FOR THIS TALK

- Theoretical Background and Testable Implications
- Data: Why Gasoline Prices?
- Previous Literature
- Methodology: Discreteness and time dependence (ACB)
- Testable Implications in the ACB Framework
- Estimation Results
- Conclusion

3 THEORIES OF PRICE ADJUSTMENT

1. Menu Cost

- A firm must pay a fixed cost in order to change its price (Dixit, 1991).
- Even though small, menu costs can exert a large impact on the business cycle (Mankiw, 1985; Fishman and Simhon, 2005).

2. Information Processing

- Processing delays (Calvo, 1983; Sims, 1998; Mankiw and Reis, 2002).
- “Inattentive producers” (Reis, 2006).
- “Inattentive consumers” (Levy, Bergen, Dutta, and Venable, 2005; Ray, Chen, Bergen, and Levy, 2006).

3. Strategic Considerations

- Customer Anger: Firms deliberately stretch out long price increases to avoid upsetting consumers (Rotemberg, 1982).
- Fairness: Firms avoid changing the price if consumers believe such a change is “unfair” (Kahneman, Knetsch, and Thaler, 1986; Rotemberg, 2002, 2006)

TESTABLE IMPLICATIONS: THE PROBABILITY OF A PRICE CHANGE

	Current price gap	Auto-correlation	History of price changes	Remaining price gap	Symmetry
Menu Costs	Yes	0	No	No	Yes
Information processing					
“Inattentive producers”					
“Inattentive consumers”					
Strategic interactions					
Partial adjustment					
Fair pricing					

TESTABLE IMPLICATIONS: THE PROBABILITY OF A PRICE CHANGE

	Current price gap	Auto-correlation	History of price changes	Remaining price gap	Symmetry
Menu Costs	Yes	0	No	No	Yes
Information processing	Yes	-	-	No	
“Inattentive producers”					Yes
“Inattentive consumers”					No (in the “small”)
Strategic interactions					
Partial adjustment					
Fair pricing					

TESTABLE IMPLICATIONS: THE PROBABILITY OF A PRICE CHANGE

	Current price gap	Auto-correlation	History of price changes	Remaining price gap	Symmetry
Menu Costs	Yes	0	No	No	Yes
Information processing	Yes	-	-	No	
“Inattentive producers”					Yes
“Inattentive consumers”					No (in the “small”)
Strategic interactions	Yes	+	+		
Partial adjustment				Yes	
Fair pricing				No	No (in the “large”)

TESTABLE IMPLICATIONS: THE PROBABILITY OF A PRICE CHANGE

	Current price gap	Auto-correlation	History of price changes	Remaining price gap	Symmetry
Menu Costs	Yes	0	No	No	Yes
Information processing	Yes	-	-	No	
“Inattentive producers”					Yes
“Inattentive consumers”					No (in the “small”)
Strategic interactions	Yes	+	+		
Partial adjustment				Yes	
Fair pricing				No	No (in the “large”)



A ROADMAP FOR THIS TALK

- Theoretical Background and Testable Implications
- **Data: Why Gasoline Prices?**
- Previous Literature
- Methodology: Discreteness and time dependence (ACB)
- Testable Implications in the ACB Framework
- Estimation Results
- Conclusion

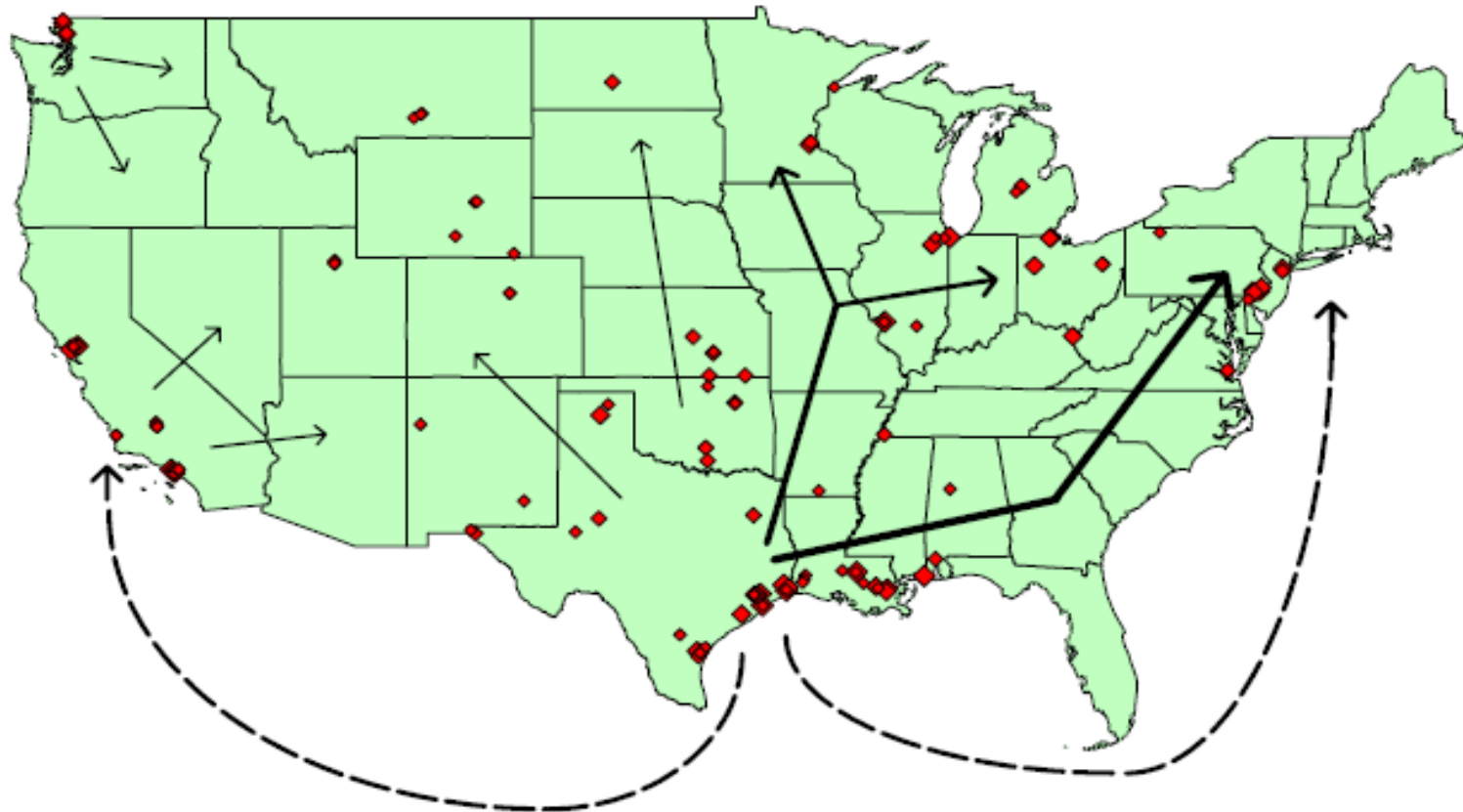


DATA - WHOLESALE GASOLINE PRICES

- Daily observations of prices charged by 9 Philadelphia gasoline wholesalers between January 1, 1989 and December 31, 1991.

- Why?
 - 5 reasons

OIL REFINERIES AND REFINED PRODUCT MOVEMENT



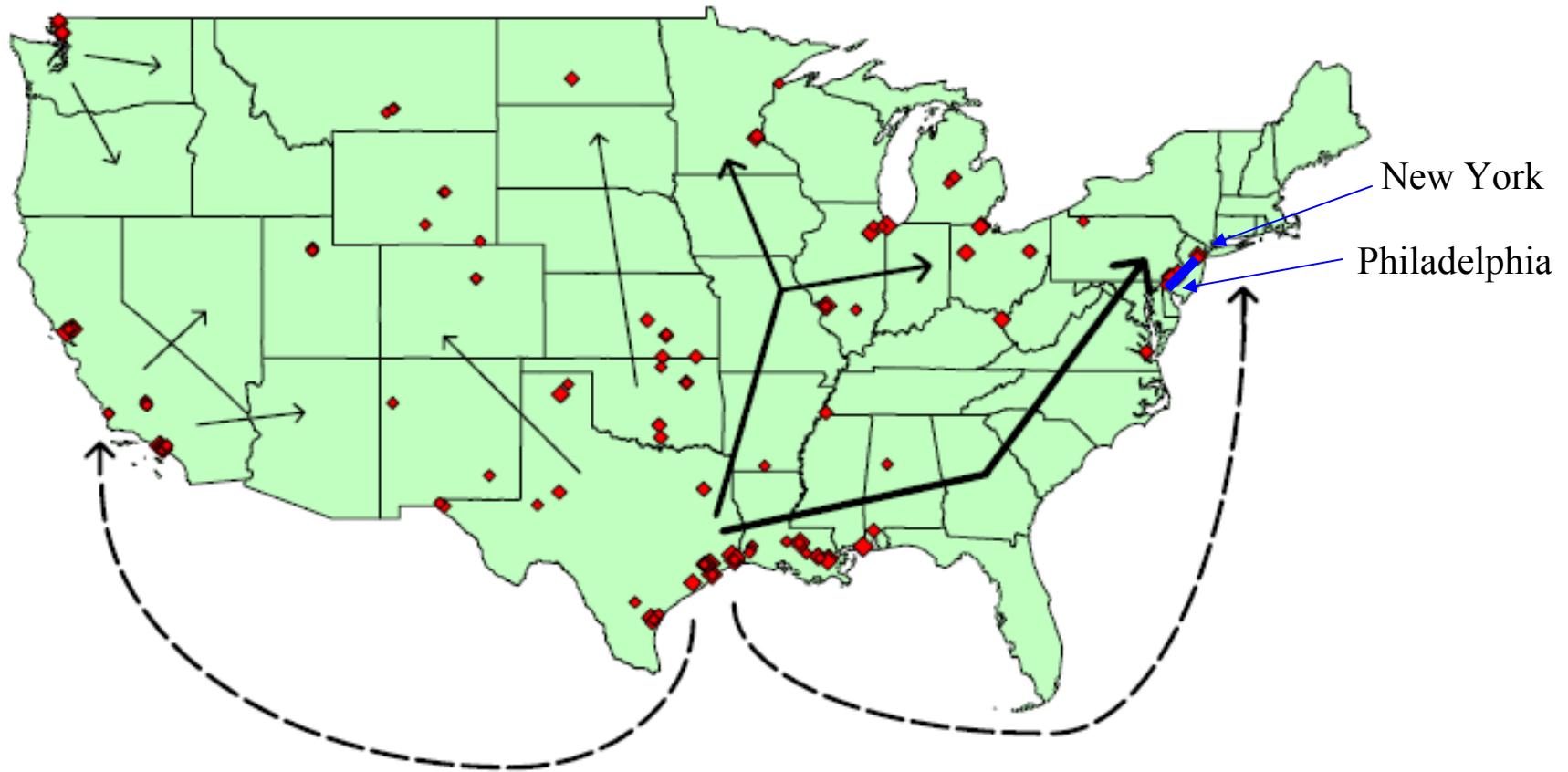
Refined Product Movements

- Pipeline
- - - Barge

Domestic Oil Refineries

- Less Than 50k bbls/day
- Between 50k and 100k bbls/day
- Between 100k and 250k bbls/day
- Over 250k bbls/day

REASON 1: MAIN INPUT COST IS EASILY OBSERVABLE



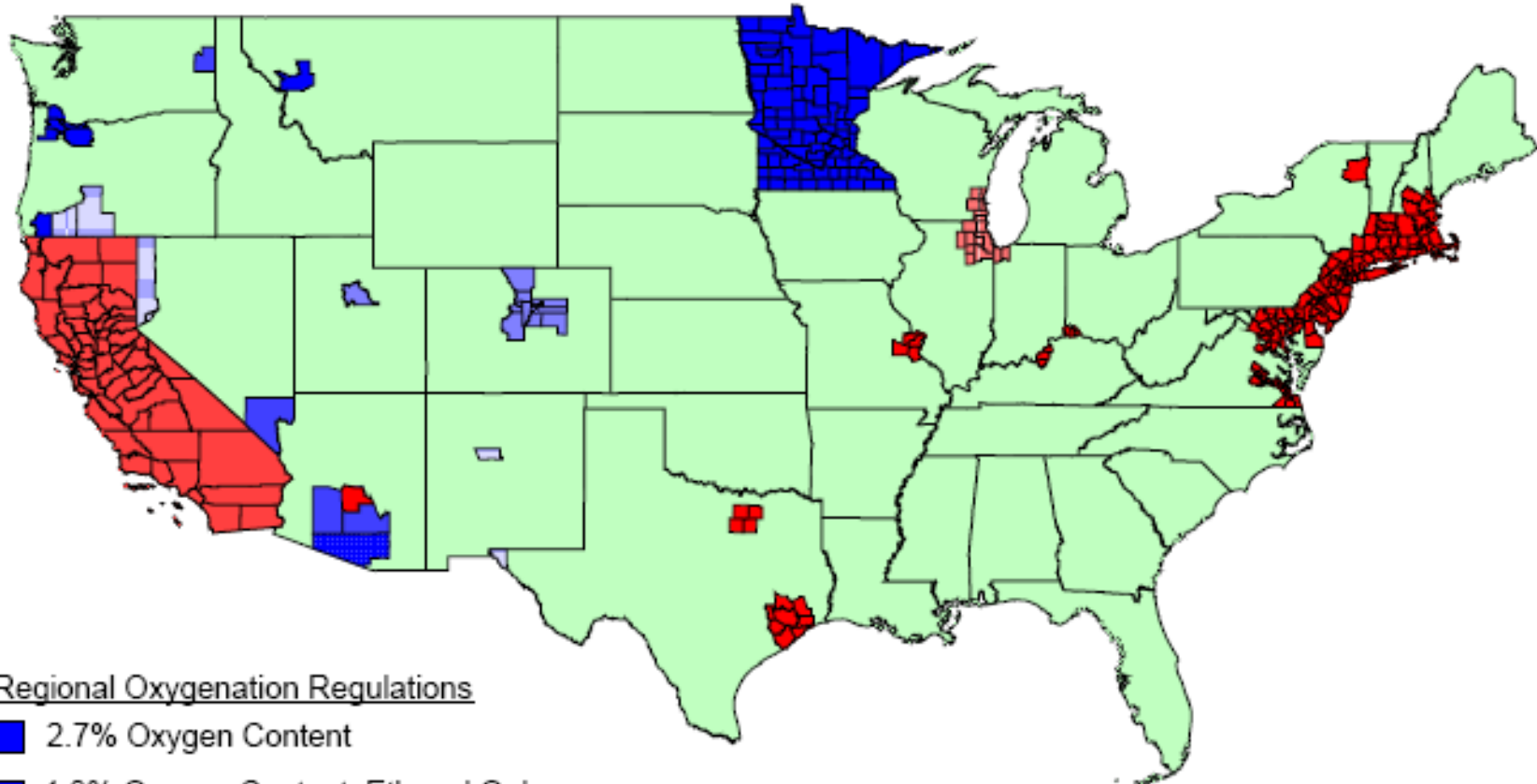
Refined Product Movements

- Pipeline
- - - Barge

Domestic Oil Refineries

- Less Than 50k bbls/day
- ◆ Between 50k and 100k bbls/day
- ◆ Between 100k and 250k bbls/day
- ◆ Over 250k bbls/day

REGIONAL GASOLINE CONTENT REGULATION



Regional Oxygenation Regulations

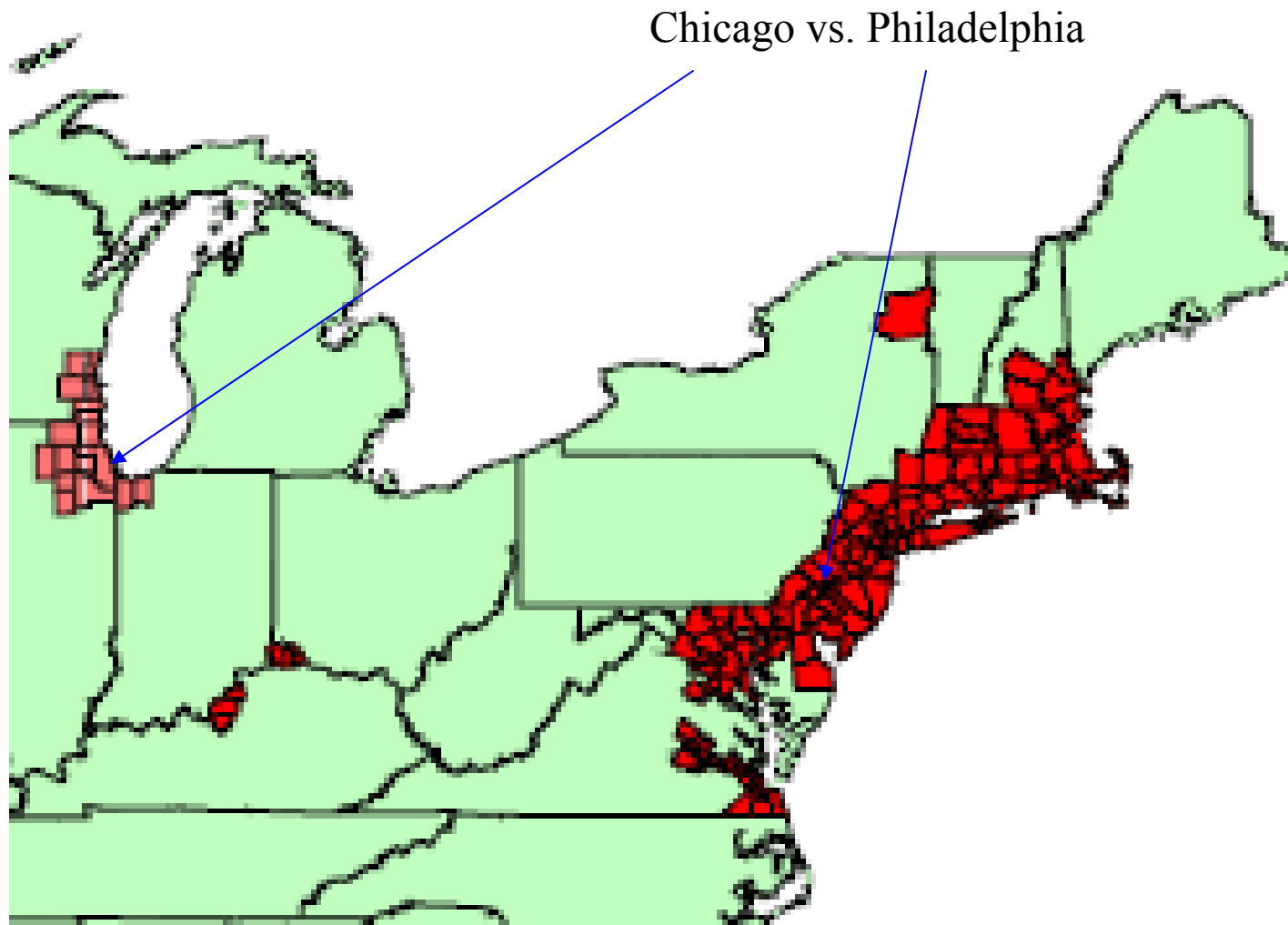
- 2.7% Oxygen Content
- 1.8% Oxygen Content, Ethanol Only
- 2.7% Oxygen Content, Ethanol Only
- 3.1% Oxygen Content, Ethanol Only
- 3.5% Oxygen Content, Ethanol Only

Regional RFG Regulations

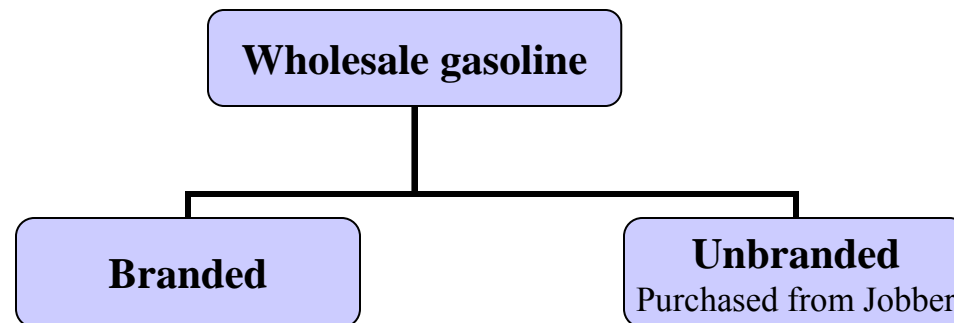
- Federal Reformulated Gasoline
- CARB Gasoline
- Ethanol-Blended RFG

Note: Minnesota mandates year-round oxygenation. Other oxygenation mandates only affect winter gasoline content.

**REASON 2:
HOMOGENEOUS GOOD**



WHOLSALER-RETAIL STATION RELATION



Company-op

Refiner owns the station
Employee of refiner operates station
Directly supplied by refiner

Lessee-dealer

Refiner owns station
Leased to third party
Mainly purchase from Jobbers

Dealer-owned

Individual retailer owns station
Under contract to sell specific brand
Purchase from Jobbers

REASON 3: PRICE CHANGES IN LIEU OF QUANTITY



Lots of 1 gallon



BUT... ARE GASOLINE PRICES STICKY?

- **Bils and Klenow (JPE, 2004)**
 - “Prices seldom change for some goods; for example, prices of newspapers, men’s haircuts, and taxi fares change less than 5 percent of months. But some prices change very frequently, with prices of **gasoline**, tomatoes, and airfares changing more than 70 percent of months.”
 - The monthly frequency of price changes ranges from 1.2 percent for coin-operated apparel laundry and dry cleaning to **79 percent for regular unleaded gasoline**.

REASON 4: STICKINESS IN WHOLESALE GASOLINE PRICE

Firm	Brand	Number of observations	Frequency of price change	Frequency of price increase	Frequency of price decrease	Average increase	Average decrease
1	Amoco	782	0.35	0.16	0.19	0.87	0.70
2	ARCO	782	0.46	0.21	0.25	0.85	0.70
3	BP	782	0.57	0.24	0.33	1.42	1.03
4	Chevron	641	0.37	0.29	0.28	0.95	0.81
5	Exxon	782	0.48	0.23	0.25	0.83	0.74
6	Gulf	743	0.41	0.20	0.21	0.87	0.70
7	Mobil	779	0.45	0.21	0.24	0.82	0.65
8	Sunoco	782	0.45	0.21	0.24	0.76	0.66
9	Texaco	681	0.40	0.19	0.21	0.90	0.66

- In contrast, the frequency of price changes for the NYMEX price of unleaded gasoline delivered at the New York Harbor was **0.95**

REASON 5: HOW TO MODEL DISCRETENESS OF PRICES AND TIME DEPENDENCE?

Firm	Number of price changes	Increase following increase	Increase following a decrease	Decrease following a increase	Decrease following a decrease
1	269	102	22	23	122
2	360	124	42	43	151
3	445	122	68	67	188
4	235	98	20	21	96
5	377	134	47	48	148
6	303	117	28	28	130
7	348	126	34	35	153
8	349	139	26	27	157
9	272	104	22	22	124

- Evidence of time dependence in data...but there is no evidence in Autoregressive Conditional Hazard Model (Davis and Hamilton, 2004)



A ROADMAP FOR THIS TALK

- Theoretical Background and Testable Implications
- Data: Why Gasoline Prices?
- **Previous Literature**
- Methodology: Discreteness and time dependence (ACB)
- Testable Implications in the ACB Framework
- Estimation Results
- Conclusion

WHY ARE GASOLINE PRICES STICKY?

THE EXISTING LITERATURE

- Industrial Organization literature: focus on “rockets and feathers”
 - Question: is there a systematic tendency for downstream prices in the oil well-to-service station gasoline industry to respond to increases in upstream prices more rapidly than downstream prices respond to decreases in upstream prices?
 - Data and Methodology:
 - Error Correction Models with or without dynamics

WHY ARE GASOLINE PRICES STICKY?

THE EXISTING LITERATURE

□ Davis and Hamilton (JMCB, 2004)

■ Dixit's menu cost model:

Firm chooses t_1, t_2, \dots to minimize

$$E_{t_0} \left\{ \sum_{i=1}^{t_i} \left[\int_{t_{i-1}}^{t_i} e^{-\rho t} k [p(t_{i-1}) - p^*(t)]^2 dt + g e^{-\rho t_i} \right] \right\}$$

■ then the probability of a price change is given by

$$h_{t+1} = h[p(t), p^*(t)] = 1 + \Phi\left(\frac{p(t) - p^*(t) - b}{\sigma}\right) - \Phi\left(\frac{p(t) - p^*(t) + b}{\sigma}\right)$$

Optimal decision rule is for the firm to change the price whenever

$$|p(t_{i-1}) - p^*(t_i)| > b = \left(\frac{6g\sigma^2}{k}\right)^{1/4}$$

DAVIS AND HAMILTON (Continued)

□ Findings:

- Dixit's model is "broadly consistent" with the data
Yet...
 - Implied menu costs are too large
 - A logit model with the same explanatory variable (price-cost gap) outperforms a structural menu cost model.

- No time dependence present in the pricing decision (Autoregressive Conditional Hazard outperformed by logit).



A ROADMAP FOR THIS TALK

- Theoretical Background and Testable Implications
- Data: Why Gasoline Prices?
- Previous Literature
- **Methodology: Discreteness and time dependence (ACB)**
- Testable Implications in the ACB Framework
- Estimation Results
- Conclusion

WHEN WILL THE NEXT PRICE CHANGE OCCUR?

A POINT PROCESS

- A point process can be described either in terms of the sequence of arrival times t_i or the sequence of durations u_i .
- Engle and Russell (1998) propose the Autoregressive Conditional Duration (ACD) to model the distribution of waiting times u_i conditional on the history of arrival times.
- Many point processes have been used in other fields of statistics

AUTOREGRESSIVE CONDITIONAL HAZARD

The ACH Model

$$h_{t+1} = \frac{1}{\psi_{N(t)} + \gamma' \mathbf{z}_t}$$

where

$$\psi_n = \alpha \sum_{i=1}^{n-1} \beta^{i-1} u_{n-i} + \beta^{n-1} \bar{u}$$

h_{t+1} : probability of a price change

$\psi_{N(t)}$: expected duration between N^{th} and $(N-1)^{th}$ price changes

u_{n-i} : observed duration

\bar{u} : average duration

ACH SPECIFICATION AND FLEXIBILITY LOSS

- Some of the flexibility of the ACD is lost by using a linear specification for the mean ψ_i
- Need to use a smoothing function

$$h_{t+1} = \frac{1}{\psi_{N(t)} + \gamma' \mathbf{z}_t}$$

- Time dependence is captured only through dependence in arrival times (durations)

AUTOREGRESSIVE CONDITIONAL BINOMIAL MODEL (ACB)

Let the response probability be given by:

$$h_t \equiv \text{prob}(x_t = 1 \mid x_t, x_{t-1}, \dots, x_1, \mathbf{z}_{t-1})$$

Then, the $ACB(q,r,s)$ model is given by:

$$G^{-1}(h_t) = \omega + \sum_{j=1}^q \alpha(x_{t-j} - h_{t-j}) + \sum_{j=1}^r \beta G^{-1}(h_{t-j}) + \sum_{j=1}^s \delta x_{t-j} + \gamma \mathbf{z}_{t-1}$$

The dynamics of the response probability are given by:

$$h_t = G \left[\omega + \sum_{j=1}^q \alpha(x_{t-j} - h_{t-j}) + \sum_{j=1}^r \beta G^{-1}(h_{t-j}) + \sum_{j=1}^s \delta x_{t-j} + \gamma \mathbf{z}_{t-1} \right]$$

THE ACB (Continued)

- Conditional on x_t , and h_t the log-likelihood can be written recursively and maximized via MLE.

- Time dynamics are captured by:
 - History of price changes
 - History of probabilities of a price change.

- The ACB(0,0,0) is a standard logit model

- *Advantages:*
 - Testing time dynamics is straight-forward.
 - Testing implications of alternative models of price stickiness is easy.

THE ACB (Continued)

- *Furthermore:* We can also investigate the **role of durations** by combining ACB with Autoregressive Conditional Duration (ACD) model.

- Use Nelson's (1991) ACD form

$$\ln(\psi_{N(t)}) = \phi + \rho \frac{u_{N(t)-1}}{\psi_{N(t)-1}} + \xi \ln(\psi_{N(t)-1})$$

- Include $\ln(u_{N(t)})$ as an explanatory variable in the ACB model
- Estimate the ACB-ACD model jointly



A ROADMAP FOR THIS TALK

- Theoretical Background and Testable Implications
- Data: Why Gasoline Prices?
- Previous Literature
- Methodology: Discreteness and time dependence (ACB)
- **Testable Implications in the ACB Framework**
- Estimation Results
- Conclusion

TESTABLE IMPLICATIONS IN THE **ACB FRAMEWORK**: THE PROBABILITY OF A PRICE CHANGE

	Current price gap $ P_t - P_t^* $	Auto- correlation $G^{-1}(h_{t-1})$	History of price changes x_{t-1}
Menu Costs	$\gamma \neq 0$	$\beta = 0$	$\delta = 0$
Information processing	$\gamma \neq 0$	$\beta < 0$	$\delta < 0$
“Inattentive producers”			
“Inattentive consumers”			
Strategic interactions	$\gamma \neq 0$	$\beta > 0$	$\delta > 0$
Partial adjustment			
Fair pricing			

$$h_t = G \left[\omega + \sum_{j=1}^r \beta G^{-1}(h_{t-j}) + \sum_{j=1}^s \delta x_{t-j} + \gamma |P_{t-1} - P_{t-1}^*| \right]$$

ALTERNATIVE SPECIFICATIONS

□ Basic specification

- $|P_t - P_t^*|$: price-cost gap. Optimal price defined as input cost plus average mark-up.

□ Additional Dynamics

- $|P_{t-1} - P_{t-1}^*|$: one day lag of gap.

□ Alternatives

- $|P_{wl(t)} - P_{wl(t)}^*|$: amount of gap remaining after previous correction, dated by $wl(t)$.
- Asymmetry: $z_t = [\theta_t, \theta_t(P_t - P_t^*), (1 - \theta_t), -(1 - \theta_t)(P_t - P_t^*)]$
where $\theta_t = 1$ if $P_t - P_t^* \geq 0$
 - For “small gaps”: $P^*(t) - P(t) \approx 0 \Rightarrow$ is the constant different?
 - For “large gaps”: $P^*(t) - P(t) \neq 0 \Rightarrow$ is the slope different?
- $u_{N(t)}$: duration between price changes

TESTABLE IMPLICATIONS IN THE **ACB FRAMEWORK**:
 THE PROBABILITY OF A PRICE CHANGE

	Current price gap $ P_t - P_t^* $	Auto-correlation $G^{-1}(h_{t-1})$	History of price changes x_{t-1}	Remaining price gap $ P_{w1(t)} - P_{w1(t)}^* $	Symmetry
Menu Costs	$\gamma \neq 0$	$\beta = 0$	$\delta = 0$	No	Yes
Information processing	$\gamma \neq 0$	$\beta < 0$	$\delta < 0$	No	
“Inattentive producers”					Yes
“Inattentive consumers”					No (in the “small”)
Strategic interactions	$\gamma \neq 0$	$\beta > 0$	$\delta > 0$		
Partial adjustment				Yes	
Fair pricing				No	No (in the “large”)



A ROADMAP FOR THIS TALK

- Theoretical Background and Testable Implications
- Data: Why Gasoline Prices?
- Previous Literature
- Methodology: Discreteness and time dependence (ACB)
- Testable Implications in the ACB Framework
- **Estimation Results**
- Conclusion

ESTIMATION RESULTS

Firm	ω	β	δ	$ P_t - P_t^* $	$ P_{t-1} - P_{t-1}^* $	log L	LR	RV
1	-1.601 (0.234)	-0.505** (0.145)	-0.184 (0.151)	-0.0364 (0.0386)	0.196** (0.0402)	-478.97	0.00070**	1.82 .0344
2	-0.110 (0.0595)	0.828** (0.117)	0.149 (0.0802)	0.107** (0.0350)	-0.102** (0.0331)	-529.43	0.0324*	0.968 .1665
3	-0.0898 (0.0987)	0.368* (0.174)	0.320* (0.140)	0.296** (0.0551)	-0.247** (0.0525)	-512.83	0.0000**	2.33 .0099
4	-0.638 (0.272)	0.467* (0.239)	0.508** (0.174)	0.106* (0.0433)	-0.0680 (0.0581)	-405.37	0.00530**	2.26 .0119
5	-0.0992 (0.0513)	0.901** (0.0464)	0.202** (0.0900)	0.113** (0.0296)	-0.114** (0.0292)	-520.52	0.0000**	1.38 .0838
6	-0.209 (0.129)	0.827** (0.105)	0.206 (0.121)	0.185** (0.0333)	-0.169** (0.0352)	-471.30	0.0183*	3.45 .0003
7	-0.0672 (0.0507)	0.899** (0.0608)	0.0686 (0.0696)	0.121** (0.0276)	-0.117** (0.0267)	-521.19	0.1136	1.59 .0559
8	-0.868 (0.253)	-0.570* (0.256)	-0.00605 (0.210)	0.0223 (0.0467)	0.126** (0.0419)	-524.31	0.0990	2.11 .0174
9	-0.267 (0.101)	0.780** (0.0798)	0.259* (0.115)	0.157** (0.0321)	-0.141** (0.0330)	-432.65	0.0186*	2.09 .0183

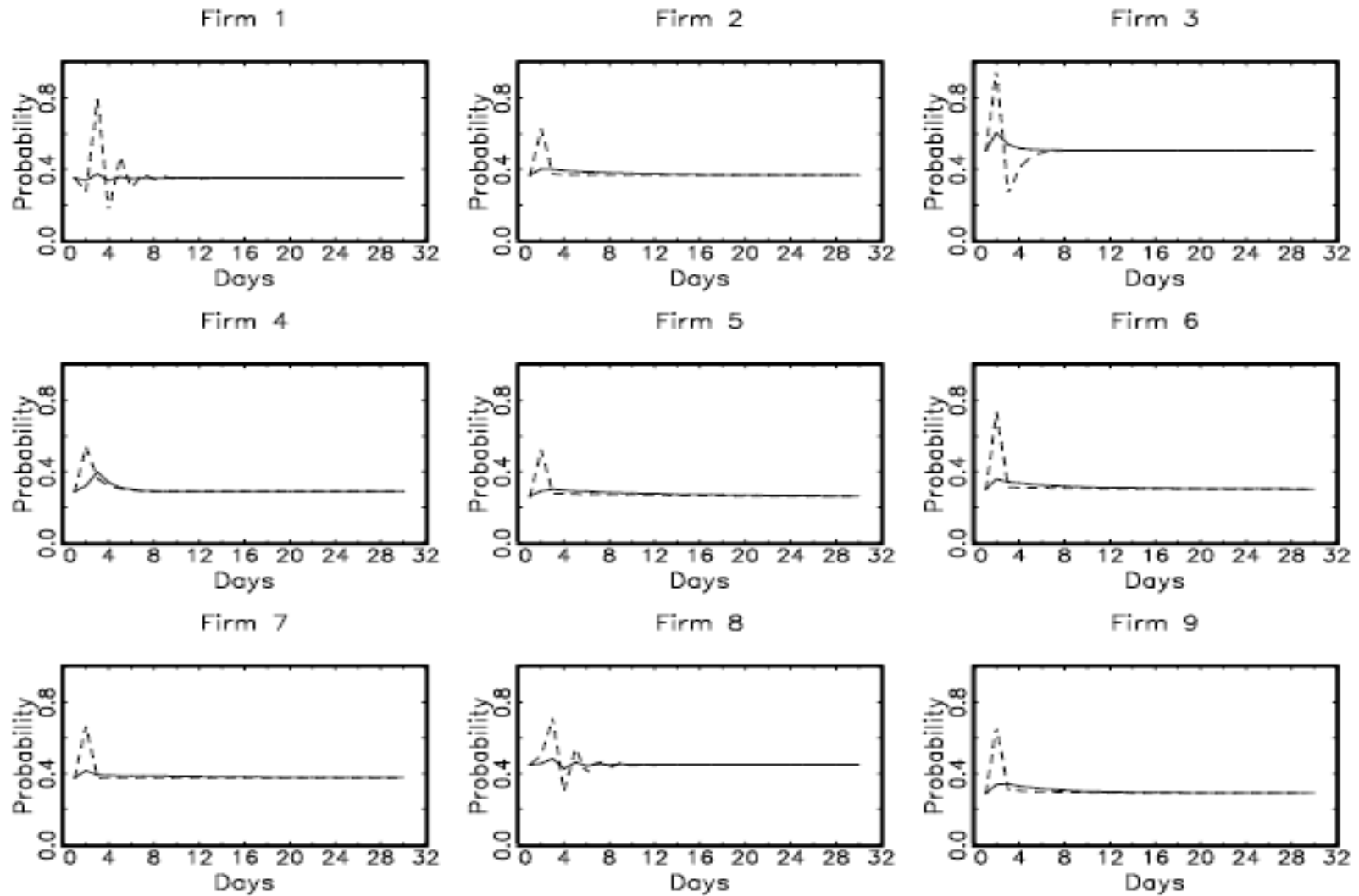
ESTIMATION RESULTS

Firm	ω	β	δ	$ P_t - P_t^* $	$P_{t-1} - P_{t-1}^*$	log L	LR
1	-1.601 (0.234)	-0.505** (0.145)	-0.184 (0.151)	-0.0364 (0.0386)	0.196** (0.0402)	-478.97	0.00070**
2	-0.110 (0.0595)	0.828** (0.117)	0.149 (0.0802)	0.107** (0.0350)	-0.102** (0.0331)	-529.43	0.0324*
3	-0.0898 (0.0987)	0.368* (0.174)	0.320* (0.140)	0.296** (0.0551)	-0.247** (0.0525)	-512.83	0.0000**
4	-0.638 (0.272)	0.467* (0.239)	0.508** (0.174)	0.106* (0.0433)	-0.0680 (0.0581)	-405.37	0.00530**
5	-0.0992 (0.0513)	0.901** (0.0464)	0.202** (0.0900)	0.113** (0.0296)	-0.114** (0.0292)	-520.52	0.0000**
6	-0.209 (0.129)	0.827** (0.105)	0.206 (0.121)	0.185** (0.0333)	-0.169** (0.0352)	-471.30	0.0183*
7	-0.0672 (0.0507)	0.899** (0.0608)	0.0686 (0.0696)	0.121** (0.0276)	-0.117** (0.0267)	-521.19	0.1136
8	-0.868 (0.253)	-0.570* (0.256)	-0.00605 (0.210)	0.0223 (0.0467)	0.126** (0.0419)	-524.31	0.0990
9	-0.267 (0.101)	0.780** (0.0798)	0.259* (0.115)	0.157** (0.0321)	-0.141** (0.0330)	-432.65	0.0186*

ESTIMATION RESULTS

Firm	ω	β	δ	$ P_t - P_t^* $	$ P_{t-1} - P_{t-1}^* $	log L	LR
1	-1.601 (0.234)	-0.505** (0.145)	-0.184 (0.151)	-0.0364 (0.0386)	0.196** (0.0402)	-478.97	0.00070**
2	-0.110 (0.0595)	0.828** (0.117)	0.149 (0.0802)	0.107** (0.0350)	-0.102** (0.0331)	-529.43	0.0324*
3	-0.0898 (0.0987)	0.368* (0.174)	0.320* (0.140)	0.296** (0.0551)	-0.247** (0.0525)	-512.83	0.0000**
4	-0.638 (0.272)	0.467* (0.239)	0.508** (0.174)	0.106* (0.0433)	-0.0680 (0.0581)	-405.37	0.00530**
5	-0.0992 (0.0513)	0.901** (0.0464)	0.202** (0.0900)	0.113** (0.0296)	-0.114** (0.0292)	-520.52	0.0000**
6	-0.209 (0.129)	0.827** (0.105)	0.206 (0.121)	0.185** (0.0333)	-0.169** (0.0352)	-471.30	0.0183*
7	-0.0672 (0.0507)	0.899** (0.0608)	0.0686 (0.0696)	0.121** (0.0276)	-0.117** (0.0267)	-521.19	0.1136
8	-0.868 (0.253)	-0.570* (0.256)	-0.00605 (0.210)	0.0223 (0.0467)	0.126** (0.0419)	-524.31	0.0990
9	-0.267 (0.101)	0.780** (0.0798)	0.259* (0.115)	0.157** (0.0321)	-0.141** (0.0330)	-432.65	0.0186*

DYNAMIC RESPONSE TO 10¢ and a 1.36 ¢ SHOCK



Dashed Line: 10¢ shock; Solid Line: 1.36¢ shock

ASYMMETRY

Firm	Pos const	Neg const	Pos gap	Neg gap	Lag pos gap	Lag neg gap
1	-1.7162** (0.2787)	-1.5107** (0.3082)	0.0120 (0.0910)	-0.0649 (0.0473)	0.2139** (0.0777)	0.1831** (0.0485)
2	-0.2568 (0.1570)	-0.0156 (0.1026)	0.1528** (0.0548)	0.1273** (0.0464)	-0.0877 (0.0573)	-0.1502** (0.0485)
3	-0.1039 (0.1278)	-0.0628 (0.1362)	0.3601** (0.0855)	0.2574** (0.0695)	-0.2704** (0.0884)	-0.2490** (0.0609)
4	-1.2550** (0.3817)	-0.6091** (0.2257)	0.1222* (0.0587)	0.0623 (0.0570)	0.0421 (0.0799)	-0.0492 (0.0623)
5	-0.1180 (0.0626)	-0.1460** (0.0566)	0.1347** (0.0441)	0.1420** (0.0386)	-0.1357** (0.0448)	-0.1419** (0.0382)
6	-1.9085** (0.3541)	-0.8220** (0.2607)	0.1689** (0.0677)	0.0700 (0.0476)	0.1567* (0.0789)	0.0212 (0.0535)
7	-0.0509 (0.0654)	-0.1281 (0.0666)	0.0826 (0.0474)	0.1587** (0.0406)	-0.0710 (0.0521)	-0.1507** (0.0392)
8	-0.8584** (0.2434)	-0.6660 (0.4345)	0.0283 (0.0570)	0.0200 (0.0502)	0.1526* (0.0688)	0.0756 (0.0900)
9	-0.3476** (0.1280)	-0.1961* (0.0948)	0.1640** (0.0495)	0.1506** (0.0400)	-0.1286* (0.0504)	-0.1445** (0.0411)

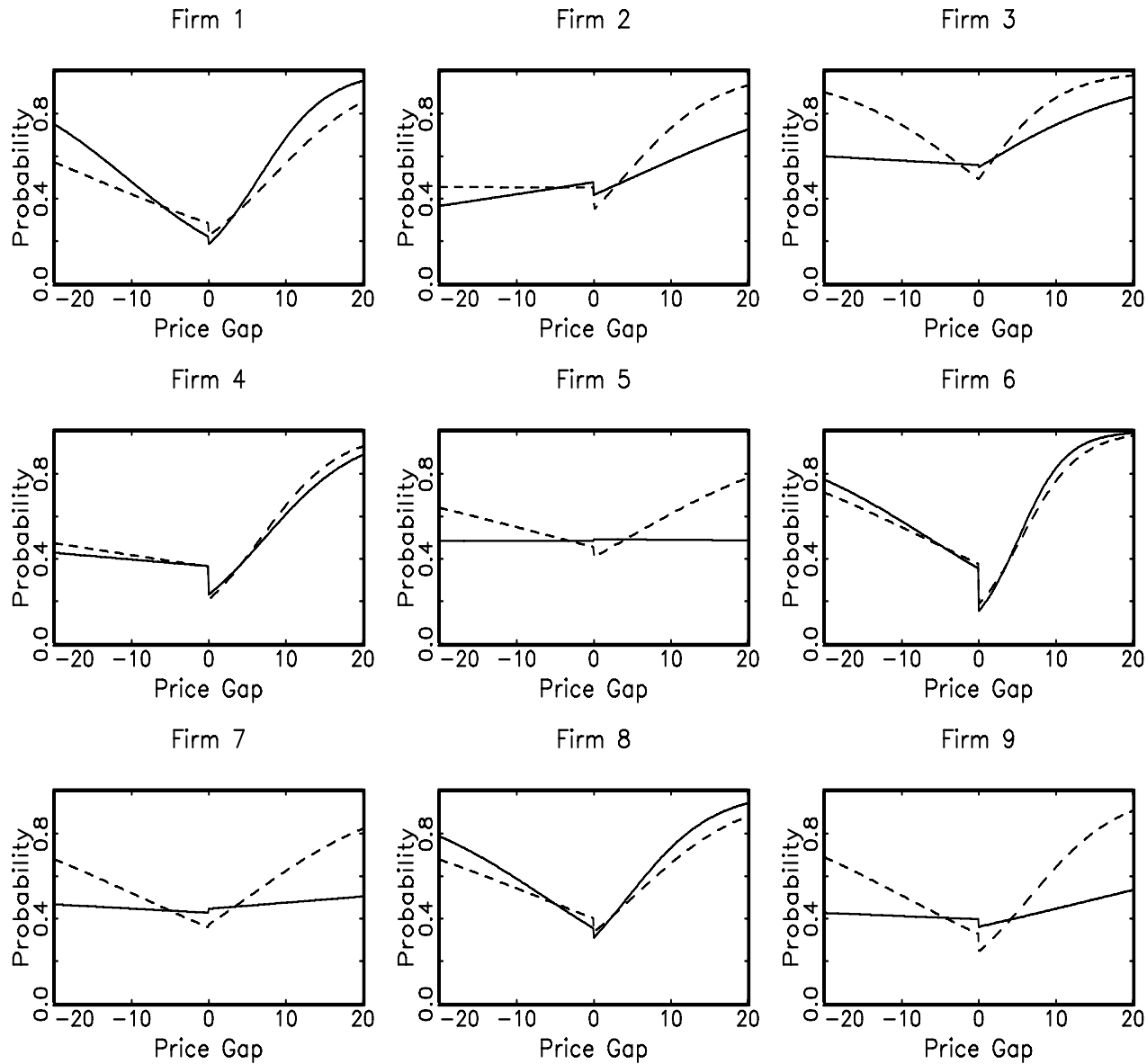
ASYMMETRY

Firm	Pos const	Neg const	Pos gap	Neg gap	Lag pos gap	Lag neg gap
1	-1.7162** (0.2787)	-1.5107** (0.3082)	0.0120 (0.0910)	-0.0649 (0.0473)	0.2139** (0.0777)	0.1831** (0.0485)
2	-0.2568 (0.1570)	-0.0156 (0.1026)	0.1528** (0.0548)	0.1273** (0.0464)	-0.0877 (0.0573)	-0.1502** (0.0485)
3	-0.1039 (0.1278)	-0.0628 (0.1362)	0.3601** (0.0855)	0.2574** (0.0695)	-0.2704** (0.0884)	-0.2490** (0.0609)
4	-1.2550** (0.3817)	-0.6091** (0.2257)	0.1222* (0.0587)	0.0623 (0.0570)	0.0421 (0.0799)	-0.0492 (0.0623)
5	-0.1180 (0.0626)	-0.1460** (0.0566)	0.1347** (0.0441)	0.1420** (0.0386)	-0.1357** (0.0448)	-0.1419** (0.0382)
6	-1.9085** (0.3541)	-0.8220** (0.2607)	0.1689** (0.0677)	0.0700 (0.0476)	0.1567* (0.0789)	0.0212 (0.0535)
7	-0.0509 (0.0654)	-0.1281 (0.0666)	0.0826 (0.0474)	0.1587** (0.0406)	-0.0710 (0.0521)	-0.1507** (0.0392)
8	-0.8584** (0.2434)	-0.6660 (0.4345)	0.0283 (0.0570)	0.0200 (0.0502)	0.1526* (0.0688)	0.0756 (0.0900)
9	-0.3476** (0.1280)	-0.1961* (0.0948)	0.1640** (0.0495)	0.1506** (0.0400)	-0.1286* (0.0504)	-0.1445** (0.0411)

ASYMMETRY

Firm	Pos const	Neg const	Pos gap	Neg gap	Lag pos gap	Lag neg gap
1	-1.7162** (0.2787)	-1.5107** (0.3082)	0.0120 (0.0910)	-0.0649 (0.0473)	0.2139** (0.0777)	0.1831** (0.0485)
2	-0.2568 (0.1570)	-0.0156 (0.1026)	0.1528** (0.0548)	0.1273** (0.0464)	-0.0877 (0.0573)	-0.1502** (0.0485)
3	-0.1039 (0.1278)	-0.0628 (0.1362)	0.3601** (0.0855)	0.2574** (0.0695)	-0.2704** (0.0884)	-0.2490** (0.0609)
4	-1.2550** (0.3817)	-0.6091** (0.2257)	0.1222* (0.0587)	0.0623 (0.0570)	0.0421 (0.0799)	-0.0492 (0.0623)
5	-0.1180 (0.0626)	-0.1460** (0.0566)	0.1347** (0.0441)	0.1420** (0.0386)	-0.1357** (0.0448)	-0.1419** (0.0382)
6	-1.9085** (0.3541)	-0.8220** (0.2607)	0.1689** (0.0677)	0.0700 (0.0476)	0.1567* (0.0789)	0.0212 (0.0535)
7	-0.0509 (0.0654)	-0.1281 (0.0666)	0.0826 (0.0474)	0.1587** (0.0406)	-0.0710 (0.0521)	-0.1507** (0.0392)
8	-0.8584** (0.2434)	-0.6660 (0.4345)	0.0283 (0.0570)	0.0200 (0.0502)	0.1526* (0.0688)	0.0756 (0.0900)
9	-0.3476** (0.1280)	-0.1961* (0.0948)	0.1640** (0.0495)	0.1506** (0.0400)	-0.1286* (0.0504)	-0.1445** (0.0411)

ASYMMETRY



THE ROLE OF DURATIONS

Firm	$\ln(u_{N(t)})$	$\ln(\tilde{u}_{N(t)-1})$	$u_{N(t-1)-1}$
1	0.0080**	0.644	0.0513
2	0.7323	0.0557	1.000
3	0.0161*	0.1797	0.1573
4	0.2404	0.1923	0.9542
5	0.1948	0.1897	0.00130***
6	0.2744	0.1512	0.2184
7	0.4074	0.5271	0.8559
8	0.2806	0.8415	1.000
9	0.7675	0.4976	0.1505

THE ROLE OF DURATIONS

Firm	$\ln(u_{N(t)})$	$\ln(u_{N(t)-1})$	$u_{N(t)-1}$
1	0.0080**	0.644	0.0513
2	0.7323	0.0557	1.000
3	0.0161*	0.1797	0.1573
4	0.2404	0.1923	0.9542
5	0.1948	0.1897	0.00130***
6	0.2744	0.1512	0.2184
7	0.4074	0.5271	0.8559
8	0.2806	0.8415	1.000
9	0.7675	0.4976	0.1505

SUMMARY OF ESTIMATION RESULTS

- Autocorrelation: $\beta > 0$ for 7 of 9 firms

- Dynamics: Cost shocks instantly passed through to retail stations.

- Asymmetry:
 - “*In the small*” for 5 out of 9 firms: More likely to raise price when gap is small and negative than lower price when gap is small and positive.

 - “*In the large*” for 6 out of 9 firms: More likely to lower price when gap is large and positive than raise price when gap is large and negative.

TESTABLE IMPLICATIONS IN THE **ACB FRAMEWORK**:
THE PROBABILITY OF A PRICE CHANGE

	Current price gap $ P_t - P_t^* $	Auto-correlation $G^{-1}(h_{t-1})$	History of price changes x_{t-1}	Remaining price gap $ P_{w1(t)} - P_{w1(t)}^* $	Symmetry
Menu Costs	$\gamma \neq 0$	$\beta = 0$	$\delta = 0$	No	Yes
Information processing	$\gamma \neq 0$	$\beta < 0$	$\delta < 0$	No	
“Inattentive producers”					Yes
“Inattentive consumers”					No (in the “small”)
Strategic interactions	$\gamma \neq 0$	$\beta > 0$	$\delta > 0$		
Partial adjustment				Yes	
Fair pricing				No	No (in the “large”)

CONCLUSION: WHY ARE PRICES STICKY?

- Results are consistent with “fair pricing”.
 - $\beta > 0$: since retailers feel entitled to their “reference transaction price”, wholesalers keep probability of price change consistent over time.
 - Cost shocks instantly passed to retailers, since they threaten wholesaler’s “reference profit”.
 - Asymmetry “in the large”, wholesalers are adverse to large upswings in price.
 - Kahneman et al (1986): Shortages will be rationed, instead of a price increase to avoid “unfair windfall”.
 - Henly, Potter, and Town (1996): Since wholesalers are tied to retailers via long term contracts, wholesalers use non-price methods of rationing in lieu of large price increases.

CONCLUSION: WHY ARE PRICES STICKY?

- Asymmetry “in the small” consistent with **rational inattention by consumers** (retailers).
 - Summary statistics: average magnitude of price increase $< \$0.01$.
 - But, retailers must change price in increments of \$0.01 or greater.
 - Thus, wholesalers have incentive to make small price increases, because they know retailers cannot follow suit.
 - Perhaps related to menu costs... but in conjunction with strategic interactions.