# Cellular service demand: biased beliefs, learning and bill shock

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#### Disclaimer

The views expressed herein are those of the authors and not necessarily those of the Bureau of Economic Analysis or the U.S. Department of Commerce.



#### Overview

- Estimate a model of tariff & usage choice, marginal-price uncertainty, biased beliefs, and learning using cellular phone billing data.
- Consumers are aware of their own uncertainty about marginal prices.
  - Incorporate optimal threshold rule for accepting/rejecting calls
- Identifying Biased Beliefs:
  - True distribution of tastes (from usage patterns)
  - Prior beliefs & learning rate (from plan choices and switches)
  - Biases measure systematic differences between the two, and lead to predictable mistakes
- Counterfactuals
  - How costly are consumer biases?
  - FCC's bill-shock regulation in 2013?



#### Data Overview

- Individual cellular billing data 8/2002 7/2004
   1366 students subscribing through a major US university.
- Pricing data for all cellular phone carriers 2002-2005 (EconOne)
- Popular plan prices, Spring 2003:

Plan	$M_j$	$Q_{j}$	$p_j$
Plan 0	\$14.99	0	\$0.11
Plan 1	\$34.99	380	\$0.45
Plan 2	\$44.99	653	\$0.40
Plan 3	\$54.99	875	\$0.40

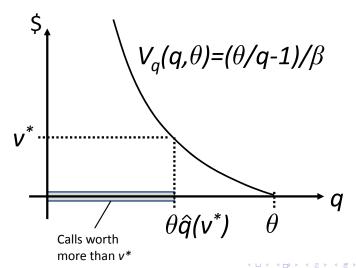
#### Illustrative Model: Timing and usage choice

- Consumers choose a plan j and a calling threshold  $v_{itj}^*$  based on beliefs about distribution of  $\theta_{it}$ .
- 2 During the course of the month  $\theta_{it}$  calling opportunities arise. Consumers make all calls worth more than  $v_{iti}^*$
- **3** Fraction  $\hat{q}(v^*)$  of calls valued above  $v^*$ . At the end of the month, realized usage is  $q_{it} = \theta_{it}\hat{q}(v_{it}^*)$ .
  - In our model  $\hat{q}(v_{it}^*) = \frac{1}{1+\beta v^*}$ .



#### Inverse Demand Curve and Calling Threshold

• Value of minutes:  $V(q, \theta) = (\theta \log(q/\theta) - q)/\beta$ 



### **Projection Bias**

• Taste shock  $\theta_{it}$  is a latent shock censored at zero

$$\theta_{it} = \left\{ \begin{array}{cc} 0 & \tilde{\theta}_{it} < 0 \\ \tilde{\theta}_{it} & \tilde{\theta}_{it} \ge 0 \end{array} \right..$$

• Latent taste shock  $\tilde{\theta}_{it} = \mu_i + \varepsilon_{it}$  is normally distributed:

Truth: 
$$ilde{ heta}_{it} \sim N\left(\mu_i, \sigma_{arepsilon}^2\right)$$

Belief: 
$$\tilde{\theta}_{it} \sim N\left(\mu_i, \tilde{\sigma}_{\varepsilon}^2\right)$$
,  $\tilde{\sigma}_{\varepsilon} = \delta_{\varepsilon} \sigma_{\varepsilon}$ 

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•  $\delta_{\varepsilon} < 1 \rightarrow$  consumers underestimate the volatility of their tastes month to month (projection bias).



#### Overconfidence

- ullet Consumers are uncertain about their own types:  $\mu_i$
- Each consumer has initial point estimate of her type  $\tilde{\mu}_{i1} \sim N(\tilde{\mu}_0, \tilde{\sigma}_{\mu}^2)$  (which she updates over time by Bayes rule.)
- Population variance of true types among customers with the same point estimate is

$$\sigma_{\mu}^2 = Var(\mu_i|\tilde{\mu}_{i1})$$

Prior beliefs:

$$\mu_i | \Im_{i,1} \sim \mathcal{N}(\tilde{\mu}_{i,1}, \tilde{\sigma}_1^2),$$

$$\tilde{\sigma}_1 = \delta_{\mu} \sigma_{\mu}$$

•  $\delta_{\mu} < 1 
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m confidence}$ , underestimate uncertainty about own type.



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Т	rue	Be	lief		Bias	Comment	
_	107	~	14		0.18	Overconfidence	
$\sigma_{\mu}$	(1.8)	$\sigma_1$	(1.0)	$o_{\mu}$	(0.01)	(too risky plans)	

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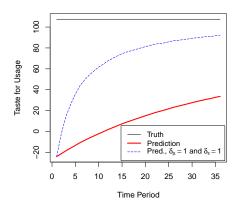
-	True	Belief		Bias	Comment
	107	<u> </u>		0.18	Overconfidence
$\sigma_{\mu}$	(1.8)	$\tilde{\sigma}_1$ (1.	0) $\delta_{\mu}$	(0.01)	(too risky plans)
$\sigma_arepsilon$	169	$\tilde{\sigma}_{arepsilon}$ (1)	$\delta_{arepsilon}$	0.54	Projection Bias
σε	(0.54)	$^{\circ\varepsilon}$ (1.	1) $^{\circ_{\varepsilon}}$	(0.01)	(too risky plans)

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	True	Bel	ief		Bias	Comment
$\sigma_{\mu}$	107 (1.8)	$ ilde{\sigma}_1$	14 (1.0)	$\delta_{\mu}$	0.18 (0.01)	Overconfidence (too risky plans)
$\sigma_{arepsilon}$	169 (0.54)	$ ilde{\sigma}_{arepsilon}$	91 (1.1)	$\delta_arepsilon$	0.54 (0.01)	Projection Bias (too risky plans)
$\mu_0$	107 (1.8)	$ ilde{\mu}_0$	-25 (6.7)	$b_1$	-132 (7.0)	Neg. Mean Bias (too small plans)

<sup>\*</sup> Risk-aversion ightarrow estimates of overconfidence and projection bias are a lower bound

### Biases Lead to Slower Learning



 Without overconfidence or projection bias, mean bias would be reduced 70% by learning in the first year. Actual reduction is only 20%.

# Impact of De-Biasing on University Plan Shares

Offered Plan	0	1	2	3	Other
Estimates	42.8	28.6	13.7	5.5	9.4
$\delta_{\mu}=1$ and $\delta_{\epsilon}=1$	39.8	25.7	17.8	7.1	9.6
No Biases	46.6	17.5	18.4	8.5	9.0



# Impact of De-Biasing on Welfare (University Plans)

	Estimates	$\delta_{\mu}=1$ and $\delta_{\epsilon}=1$	
Avg. Bill	41.37	37.96	
Ovg. Prob.	0.2	0.1	
$\Delta$ Monthly Fee		1.53	
$\Delta$ Overage Fee		-4.32	
Δ Bill		-3.41	
$\Delta$ q $\mid$ Overage		-34	
Δq		-14.5	
$\Delta$ Profit (Annual)		-40.89	
$\Delta$ Cons. Welf. (Annual)		29.9	
$\Delta$ Tot. Welf. (Annual)		-10.99	



# Impact of De-Biasing on Welfare (University Plans)

	Estimates	$\delta_{\mu}=1$ and $\delta_{\epsilon}=1$	No Biases
Avg. Bill	41.37	37.96	36.3
Ovg. Prob.	0.2	0.1	0.05
$\Delta$ Monthly Fee		1.53	0.51
$\Delta$ Overage Fee		-4.32	-5.76
Δ Bill		-3.41	-5.07
$\Delta$ q $\mid$ Overage		-34	-12
Δq		-14.5	-12.2
$\Delta$ Profit (Annual)		-40.89	-60.79
$\Delta$ Cons. Welf. (Annual)		29.9	51.3
$\Delta$ Tot. Welf. (Annual)		-10.99	-9.49



## Impact of De-Biasing on Welfare (Public Plans)

	Estimates	$\delta_{\mu}=1$ and $\delta_{\epsilon}=1$	
Avg. Bill	49.24	41.77	
Ovg. Prob.	0.23	0.12	
△ Monthly Fee		1.21	
$\Delta$ Overage Fee		-8.68	
∆ Bill		-7.47	
$\Delta$ q $\mid$ Overage		-41	
Δq		-28.1	
$\Delta$ Profit (Annual)		-89.66	
$\Delta$ Cons. Welf. (Annual)	64.83		
$\Delta$ Tot. Welf. (Annual)		-24.84	



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## Impact of De-Biasing on Welfare (Public Plans)

	Estimates	$\delta_{\mu}=1$ and $\delta_{\epsilon}=1$	No Biases
Avg. Bill	49.24	41.77	40.73
Ovg. Prob.	0.23	0.12	0.06
$\Delta$ Monthly Fee		1.21	3.18
$\Delta$ Overage Fee		-8.68	-11.69
Δ Bill		-7.47	-8.51
$\Delta$ q $\mid$ Overage		-41	-10
Δq		-28.1	-20.4
$\Delta$ Profit (Annual)		-89.66	-102.12
$\Delta$ Cons. Welf. (Annual)		64.83	86.64
Δ Tot. Welf. (Annual)		-24.84	-15.48



### Equilibrium Price Response to Bill-Shock Regulation

			Est, Bill Shock	
		Est	(fixed prices)	Est, Bill Shock
		(1)	(2)	(3)
Plan 1	М	28.46	28.46	28.36
	Q	0	0	0
	р	50	50	50
	Share	63	63	72
Plan 2	М	61.28	61.28	73.99
	Q	295	295	374
	р	50	50	50
	Share	37	37	28
Pro	ofit	915	882	919
Cons V	Velfare	5497	5515	5465
Total V	Velfare	6413	6396	6384



### Equilibrium Price Response to Bill-Shock Regulation

		$\delta_{\mu}=1$	
		and $\delta_arepsilon=1$	No Biases
		(4)	(5)
Plan 1	М	29.26	78.07
	Q	0	$\infty$
	р	50	N/A
	Share	64	49
Plan 2	Μ	78.95	78.07
	Q	$\infty$	$\infty$
	р	N/A	N/A
	Share	36	51
Pro	ofit	925	937
Cons V	Velfare	5501	5695
Total V	Velfare	6425	6632



#### Conclusion

- We estimate a model of tariff & usage choice, marginal-price uncertainty, biased beliefs, and learning using cellular phone billing data
- Estimates
  - Overconfidence: underestimate uncertainty about mean usage by 82%
  - Projection Bias: underestimate monthly volatility in usage by 46%
- Biases significantly decrease consumer welfare
  - Overconfidence and projection bias hurt consumers \$30/year (6% avg. bill)
  - All biases hurt consumers \$51/year (10% avg. bill)

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- We estimate a model of tariff & usage choice, marginal-price uncertainty, biased beliefs, and learning using cellular phone billing data
- Estimates
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  - Overconfidence and projection bias hurt consumers \$30/year (6% avg. bill)
  - All biases hurt consumers \$51/year (10% avg. bill)
- Bill shock regulation...
  - helps consumers if prices are fixed
  - may hurt consumers if prices vary



### Predictable Mistakes and Savings Opportunities

- Customers beliefs are biased, so make predictable mistakes.
- Evidence from "arbitrage opportunities"
  - University acts as a reseller, charging a fixed \$5 fee per month
  - University could bill students for their chosen plan, but sign them up for an alternative plan, and pocket the difference in charges.

	First Opportunity	Second Opportunity
Dates	10/02-8/03	9/03 onwards
Enrollment Change	plan 1-3 $ ightarrow$ plan 0	plan $1 o$ plan $2$
Affected Customers	251 (34%)	445 (55%)
Additional Revenue		
Total	\$20,840 (47%)	\$7,942 (28%)
Per Affected Bill	\$8.76	\$2.64
Per Affected Cust.	\$83.03 (149%)	\$17.85 (46%)

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# Debiasing Counterfactuals: (with University Plans)

Table: Dollar values in percentage terms

Estimates	$\delta_{\mu}=1$ and $\delta_{\epsilon}=1$	No Biases
	0.04	0.01
	-0.1	-0.14
	-0.08	-0.12
	-0.08	-0.12
	0.06	0.1
	-0.02	-0.02
	Estimates	0.04 -0.1 -0.08 -0.08

### Debiasing Counterfactuals: Public Plan Shares

Offered Plan	0	1	2	3	Other
NA	0.0	23.2	11.7	4.9	60.2
NA	0.0	20.9	13.9	5.6	59.6
NA	0.0	15.5	16.6	8.9	59.0

### Illustrative Model: Consumer Utility

- Each date t, consumer i chooses a plan j, and then a quantity  $q_{it}$ .
- Consumer utility is

$$u_{itj} = V(q_{it}, \theta_{it}) - \alpha P_j(q_{it}) + \eta_{itj}$$

 $V\left(q_{it}, \theta_{it}\right)$  is the value of consuming  $q_{it}$  units given taste shock  $\theta_{it}$ ,

$$V\left(q_{it}, \theta_{it}
ight) = rac{1}{\gamma} \left( heta_{it} ln(q_{it}/ heta_{it}) - q_{it} 
ight)$$

 $P_{j}(q_{it})$  is the payment for usage  $q_{it}$  on plan j,

$$P_{j}\left(q_{it}\right) = M_{j} + p_{j} \max\left\{0, q_{it} - Q_{j}\right\}$$

and  $\eta_{itj}$  is an iid logit error.



#### Illustrative Model: Consumer Demand

• Let  $q(p, \theta_{it})$  be consumer demand given constant marginal price p:

$$q(p, \theta_{it}) \equiv \arg\max_{q} V(q, \theta_{it}) - pq,$$

• Define  $\beta = \alpha \gamma$ . Then

$$q(p, \theta_{it}) = \theta_{it}\hat{q}(p)$$
  
 $\hat{q}(p) = 1/(1+\beta p)$ 

#### Interpretation:

- ullet  $heta_{it}$  call opportunities arise in billing period
- $\hat{q}(p)$  is the fraction of calls worth more than p



### Illustrative Model: Timing and usage choice

- Consumers choose a plan j and a calling threshold  $v_{itj}^*$  based on beliefs about distribution of  $\theta_{it}$ .
- ② During the course of the month consumers do not track usage, but simply make all calls valued above  $v_{it}^*$ :1

$$v_{itj}^* = p_j \Pr\left(\theta_{it} \ge Q_j/\hat{q}(v_{itj}^*)\right) \frac{E\left[\theta_{it} \mid \theta_{it} \ge Q_j/\hat{q}(v_{itj}^*); \ \Im_{it}\right]}{E\left[\theta_{it} \mid \Im_{it}\right]}$$

**3** At the end of the month, realized usage is  $q_{it} = \theta_{it} \hat{q}(v_{it}^*)$ .

¹Optimal strategy for an inattentive consumer who does not keep track of past usage.

#### Identification Overview

- **1** True tastes:  $\sigma_{\varepsilon}$ , and population distribution of  $\mu_i$ .
  - usage patterns
- **②** Beliefs:  $\tilde{\sigma}_{\varepsilon}$ ,  $\tilde{\sigma}_{1}$ , and population distribution of  $\tilde{\mu}_{i1}$ .
  - Initial plan choice shares and switching
- **3** Price coefficient  $\beta$ 
  - 9pm usage increase

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