

# Antitrust Contests

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# Roadmap

- › Theoretical foundations of structural estimation
  - › Overview of antitrust contests in the US
  - › Data
  - › Structural estimates
  - › Monte Carlo results: Reliability & bias
  - › Conclusions
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# Theoretical Foundations

- › Two contestants
    - £ FTC (player 1)
    - £ Defendant (player 2)
  - › Potentially different values to each of winning and losing
  - › Contest Success Function:
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# Model 1 (Generalized Tullock)

$$\Pr(\text{FTC Wins} | X) = \frac{\sigma x_1^r}{\sigma x_1^r + x_2^r}$$

$$= \frac{\sigma x_1^r}{\sigma x_1^r + x_2^r}$$

$$= \frac{1}{1 + \frac{1}{\sigma} \left(\frac{x_2}{x_1}\right)^r}$$

$$\frac{1}{1 + \frac{1}{\sigma} z^r} \equiv \frac{1}{1 + z^r}$$

## Model 2 (Logistic)

$$P(Y=1|X) = F(\omega)$$

where  $F$  is logistic:

$$F(\omega) = \frac{\exp(\omega)}{1 + \exp(\omega)}$$

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# Pros and Cons

- › Advantage of Generalized Tullock
    - £ Well established theoretical literature (by all of you and others)
  - › Advantages of Logistic
    - £ Structural micro foundations (McFadden and others)
    - £ Empirically estimable using standard logit estimation rather than problematic binomial MLE methods
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# Key Result: Structural Equivalence

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## Remark 2

- Structural equivalence works for more general contest success functions, such as

$$p(x_1, x_2) = \frac{\alpha_1 x_1^r}{\alpha_1 x_1^r + \alpha_2 \left(\frac{x_2}{x_1}\right)^r} = \frac{1}{1 + \frac{\alpha_2}{\alpha_1} \left(\frac{x_2}{x_1}\right)^r} \equiv \frac{1}{1 + \rho z^r}$$

£ But cannot separately identify  $\alpha$ 's, only the ratio

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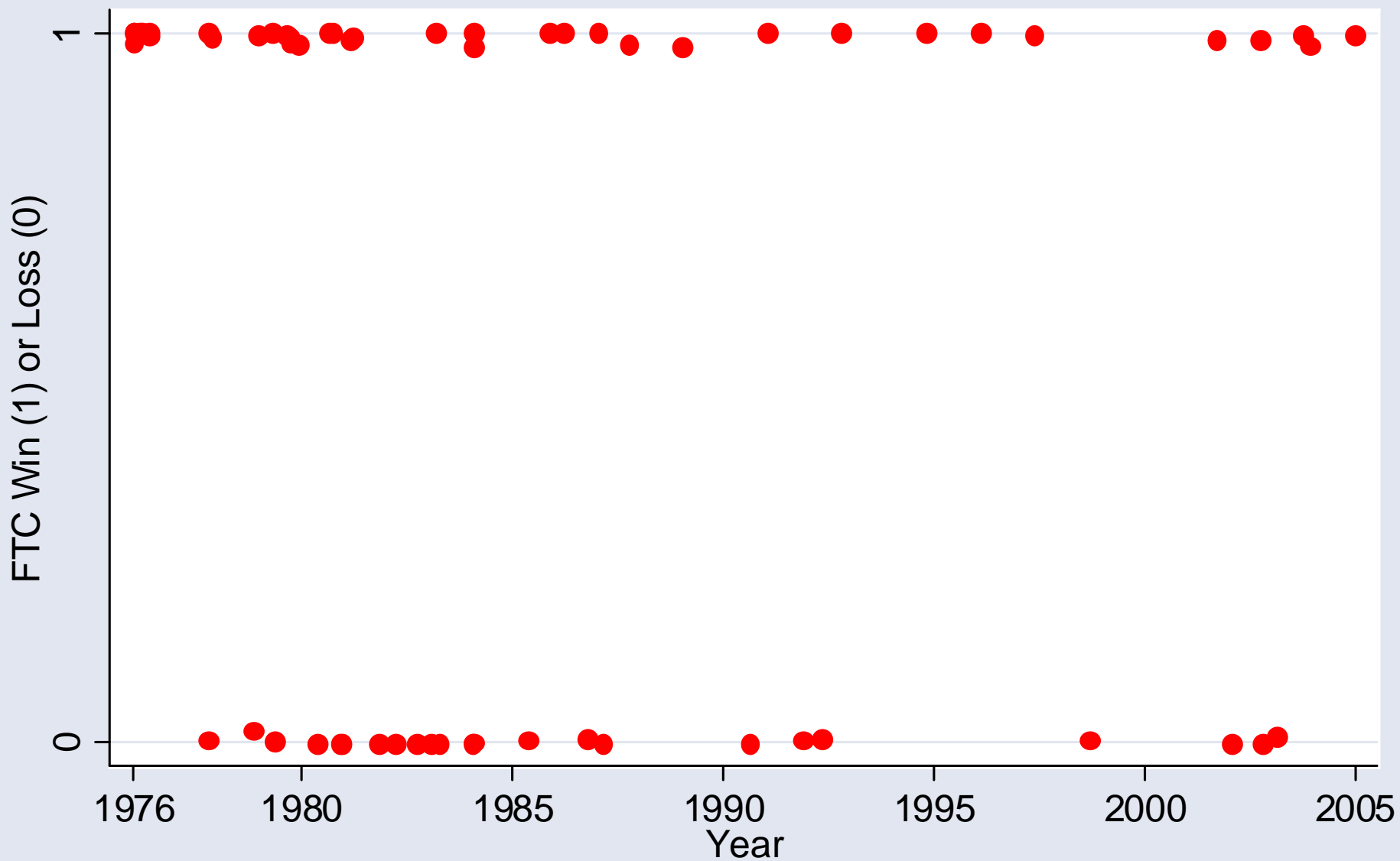
## Remark 3

- › Not generally feasible to exploit additional structure
    - £ For  $\epsilon > 0, r > 0$  equilibrium in mixed-strategies guaranteed, but structure of strategies generally unknown except for specific values of  $\epsilon$  and  $r$
    - £ For some parameter configurations, equilibrium is in pure strategies, but these regions depend on  $r, \epsilon$ , as well as the (*unknown*) values of winning and losing
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# FTC Wins and Losses

ALJ Decisions





# Results: The Good

Table 2: Implied Structural Parameters

	Baseline		Random Effects	
	Case	Other Case	Case	Other Case
$r$	0.26	0.31	0.27	0.33
$\sigma$	0.74	0.78	0.75	0.79
$\sigma$ (Order)				
$\sigma$ (Merge Case)	0.49		0.51	
$\sigma$ (Other Case)				0.51
$\sigma$ (Other Case)				0.79

# Results: The Good

**Table 2: Implied Structural Parameters**

	Baseline		Random Effects			
	(1) Baseline	(2) Controls for Type of Case	(3) Unobserved Case Heterogeneity	(4) Unobserved Case Heterogeneity	(5) Unobserved ALJ Heterogeneity	(6) Unobserved ALJ Heterogeneity
r	0.26	0.31	0.27	0.33	0.26	0.31
<b>Sigma (Pooled)</b>	<b>0.76</b>		<b>0.74</b>		<b>0.75</b>	
Sigma (Merger Case)		0.51		0.49		0.51
Sigma (Other Case)		0.79		0.78		0.79



# Results: The Good

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Sigma (Pooled)	0.76		0.74		0.75	
<b>Sigma (Merger Case)</b>		<b>0.51</b>		<b>0.49</b>		<b>0.51</b>
Sigma (Other Case)		0.79		0.78		0.79

# Results: The Good

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r	0.26	0.31	0.27	0.33	0.26	0.31
Sigma (Pooled)	0.76		0.74		0.75	
Sigma (Merger Case)		0.51		0.49		0.51
Sigma (Other Case)		0.79		0.78		0.79

# Results: The Bad

Table 1: Results of Logistic Structural Estimation

	Baseline		Random Effects			
	(1) Baseline	(2) Controls for Type of Case	(3) Unobserved Case Heterogeneity	(4) Unobserved Case Heterogeneity	(5) Unobserved ALJ Heterogeneity	(6) Unobserved ALJ Heterogeneity
LN(z)	<b>-0.256</b> (0.72)	<b>-0.309</b> (0.84)	<b>-0.269</b> (0.65)	<b>-0.327</b> (0.78)	<b>-0.258</b> (0.68)	<b>-0.311</b> (0.80)
MERGER DUMMY		-0.435 (0.79)		-0.458 (0.77)		-0.432 (0.79)
CONSTANT	-0.281 (0.26)	-0.231 (0.21)	-0.297 (0.25)	-0.247 (0.20)	-0.287 (0.25)	-0.238 (0.20)

# Results: The Bad

Table 1: Results of Logistic Structural Estimation

	Baseline		Random Effects			
	(1) Baseline	(2) Controls for Type	(3) Unobserved Case	(4) Unobserved Case	(5) Unobserved ALJ	(6) Unobserved ALJ
Number of Cases	-0.256	-0.309	0.269	0.327	0.258	0.311
Number of ALJ	-0.448	-0.448	-0.439	-0.439	-0.439	-0.439
Number of Cases	(0.256)	(0.309)	(0.269)	(0.327)	(0.258)	(0.311)
Number of ALJ	(0.448)	(0.448)	(0.439)	(0.439)	(0.439)	(0.439)
Number of Cases	(0.256)	(0.309)	(0.269)	(0.327)	(0.258)	(0.311)
Number of ALJ	(0.448)	(0.448)	(0.439)	(0.439)	(0.439)	(0.439)
Number of Cases	(0.256)	(0.309)	(0.269)	(0.327)	(0.258)	(0.311)
Number of ALJ	(0.448)	(0.448)	(0.439)	(0.439)	(0.439)	(0.439)

# Results: The Bad

Table 1: Results of Logistic Structural Estimation

	Model 1	Model 2	Model 3	Model 4	Model 5	Model 6	
LN(z)	-0.327 (0.78)	-0.258 (0.68)	-0.311 (0.80)		-0.256 (0.72)	-0.309 (0.81)	-0.269 (0.65)
LN(z)	<b>-0.458</b> (0.67)		<b>-0.432</b> (0.69)			<b>-0.435</b> (0.70)	
CONSTANT	0.267 (0.73)	0.267 (0.73)	0.267 (0.73)		0.267 (0.73)	0.267 (0.73)	0.267 (0.73)

Note: Robust z statistics in parentheses

# Results: The Bad

Table 1: Results of Logistic Structural Estimation

	Random Effects			Baseline		
	Unrestricted	Unrestricted	Unrestricted	Unrestricted	Unrestricted	Unrestricted
	0.255	0.200	0.260	0.227	0.259	0.241 (NS)
<b>Y</b>	-0.435		-0.458		-0.432	<b>MERGER DUMMI</b>
	(0.77)		(0.79)		(0.79)	
	-0.297	-0.247	-0.287	-0.238	<b>CONSTANT</b>	-0.281
	(0.25)	(0.20)	(0.25)	(0.20)		(0.26)
	60	60	60	60	<b>Observations</b>	60
			17	17	<b>Number of ALJ's</b>	60
	60	60			<b>Number of Cases</b>	

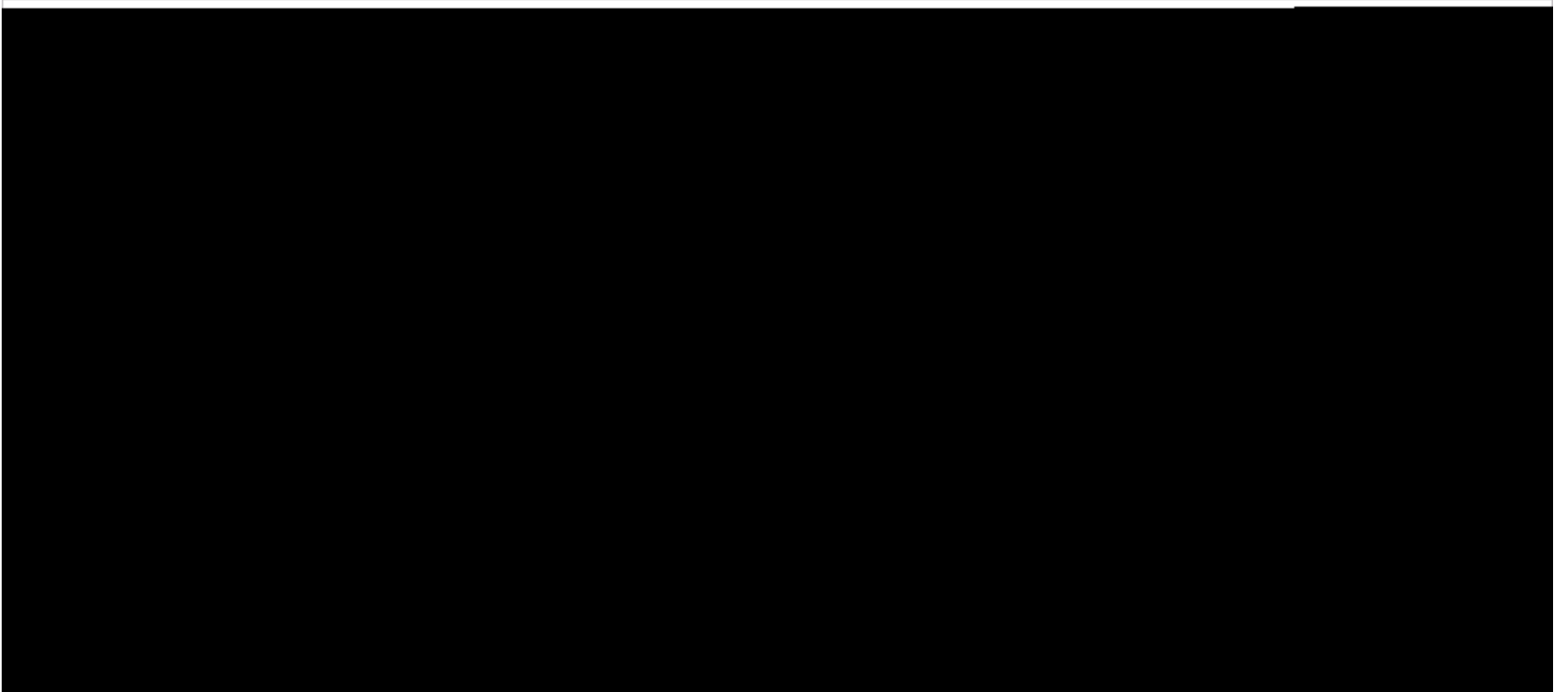
# Results: The Bad

Table 1: Results of Logistic Structural Estimation

	Baseline			Random Effects			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Dependent Variable	Class			Class			Baseline
Control	- heterogeneity			- heterogeneity			
Type of Case	- heterogeneity			- heterogeneity			
Constant	-0.309	-0.269	-0.327	-0.258	-0.311	-0.251	
Standard Error	(0.84)	(0.65)	(0.78)	(0.68)	(0.80)	(0.72)	
Number of Observations	-0.435	-0.458	-0.458	-0.432	MERGER DUMMY		
Number of Firms	-0.79	-0.77	-0.77	-0.79			
Number of Acquisitions							
Number of Alliances							
Number of Cases							
Model fit statistics							

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# Results: The Bad



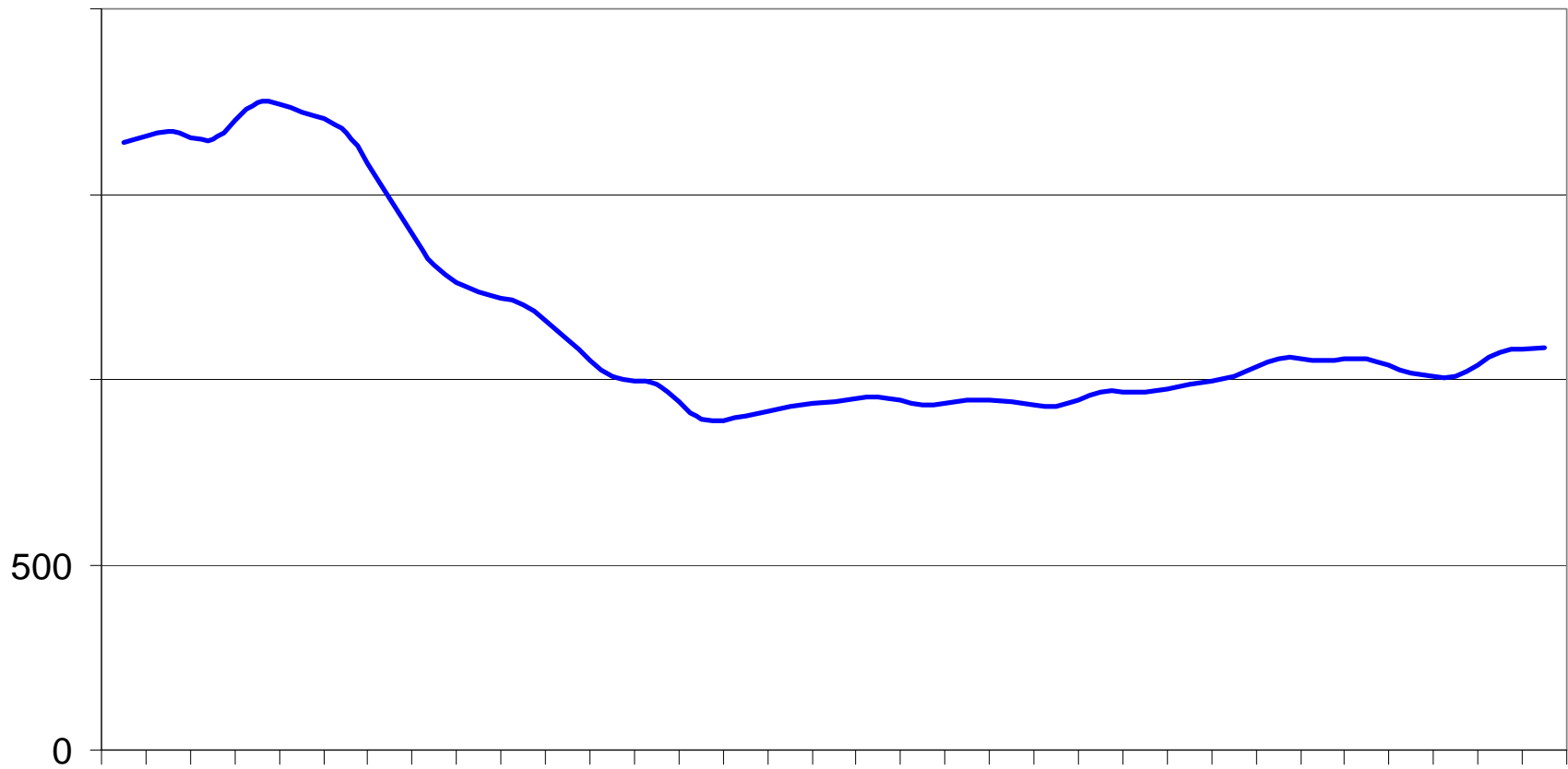


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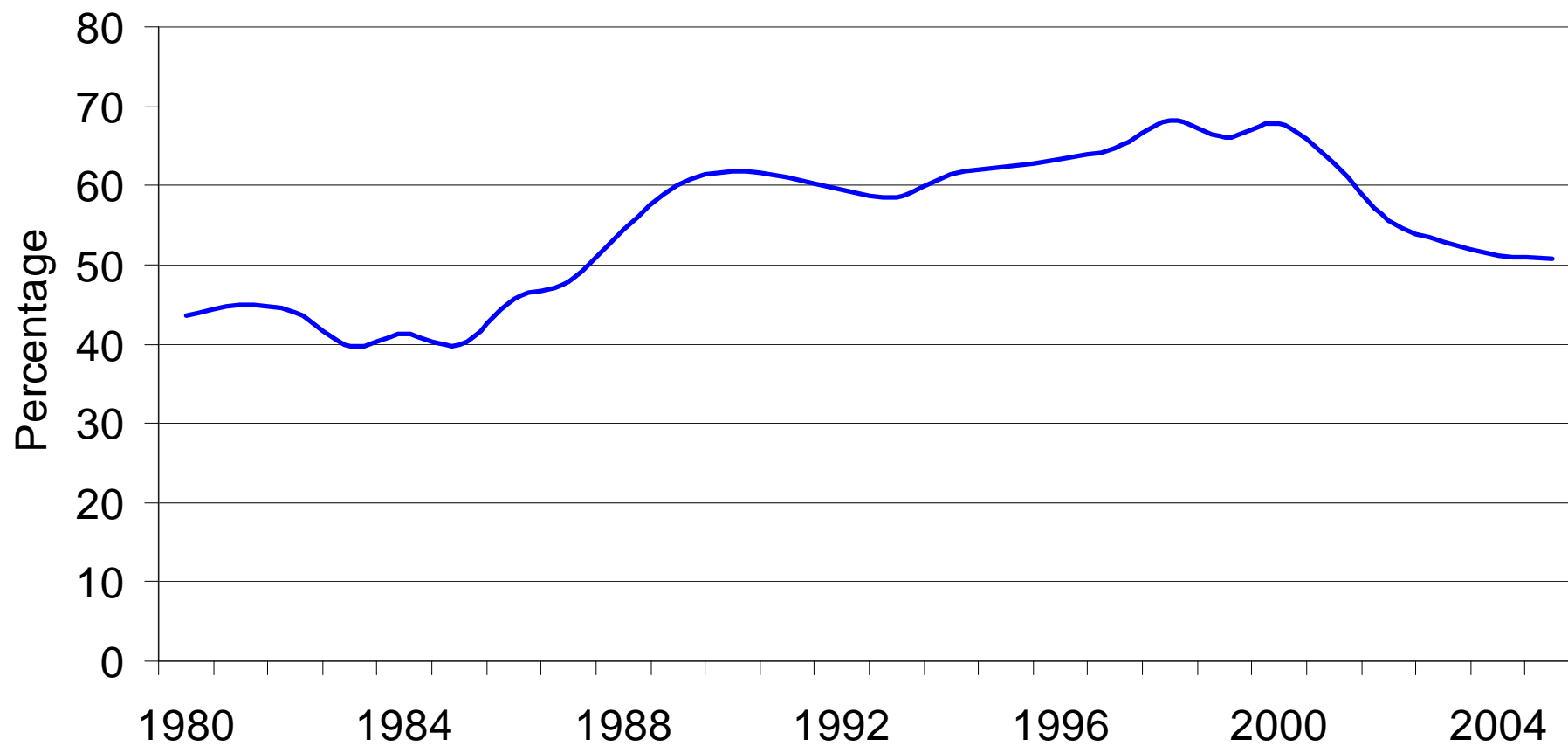
# Better Data?

- › Include other inputs (attorneys)
  - › Adjustments for time on antitrust versus other activities (consumer protection or advocacy)
  - › Expenditures on experts
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# Total FTC Employees

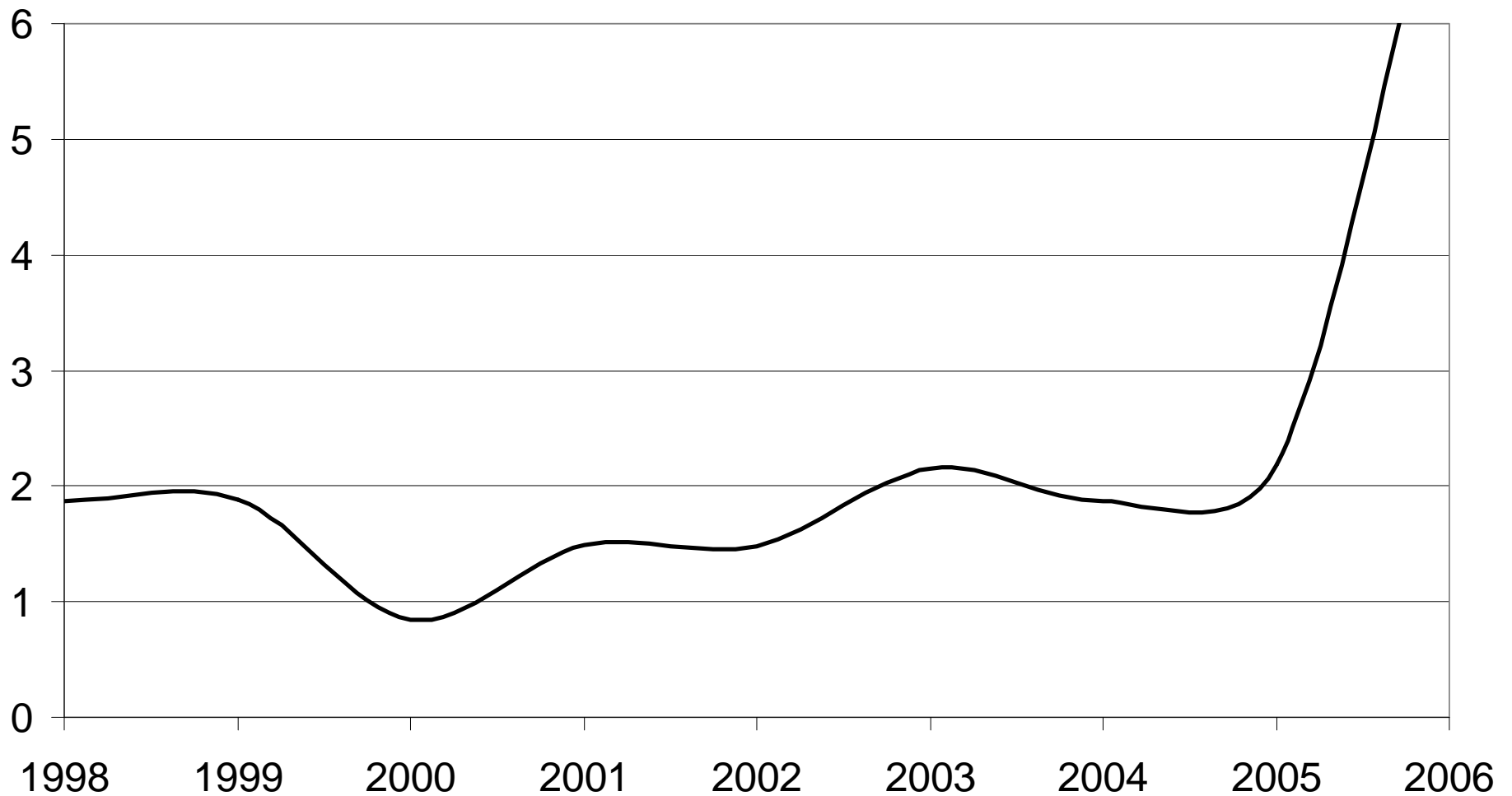


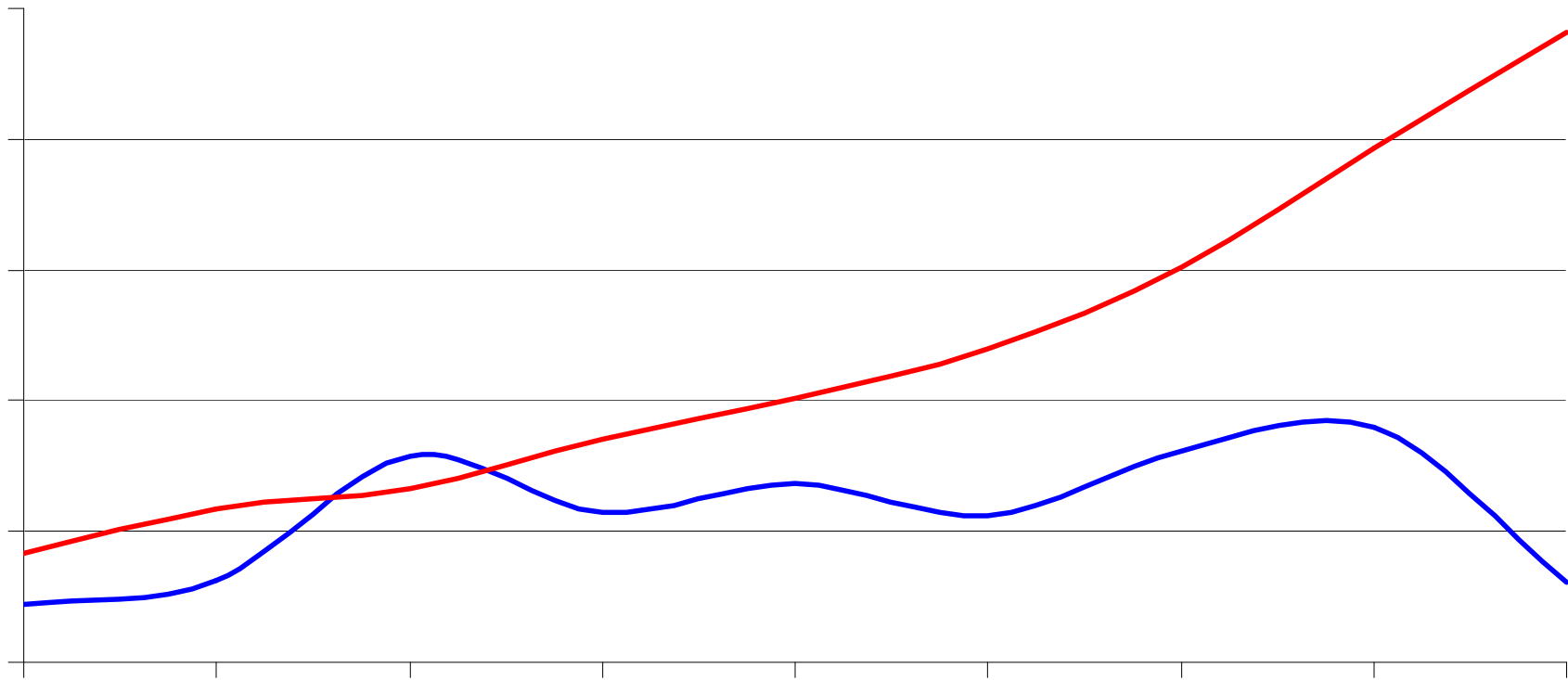
## Percentage Allocation of Economist Time for Antitrust



# Example of Alternative $z$ Measure:

**Estimated Defendant Expenditures on Economic Experts Relative to that of the FTC**





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# Results From These Data

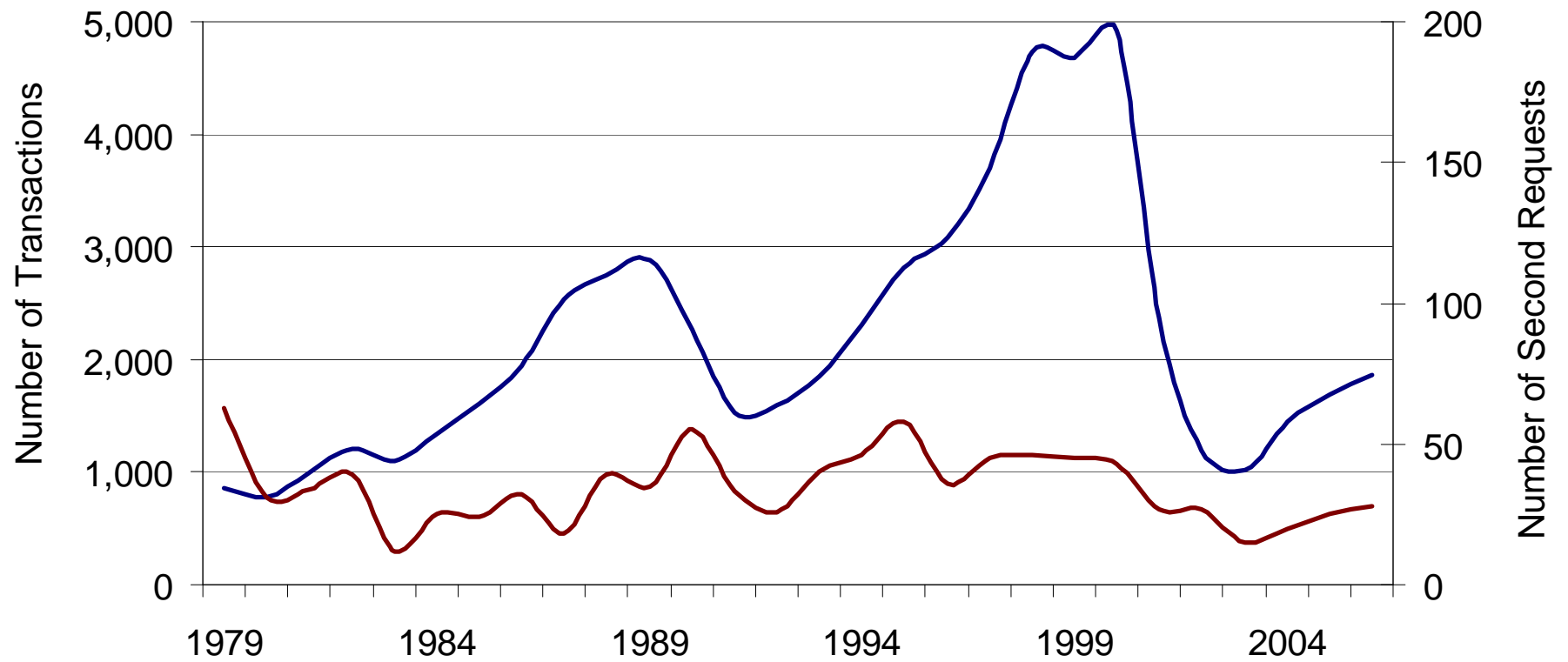
- › Similar sorts of estimates
  - › No more reliable
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# What About Endogeneity?

- › Merger activity
  - › Selection issues
  - › Endogenous effort
    - £ Impose restrictions on  $z$  implied by PSNE and use proxies for values of winning
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# Hart-Scott-Rodino Transactions & Second Requests



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# Accounting for Endogeneity

- › Doesn't help!
- › What's going on?



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# Monte Carlo

- › Generated data from a “true” model

$$r \in \{.25, 1, 1.5\}$$

$$\sigma \in \{.25, 1, 1.5\}$$

- › Low, medium, high cross sectional variation in  $z$   
(measured by coefficient of variation)
  - › 20 obs, 60 obs, 400 obs
  - › Replicated 10,000 times each
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Table 5: Monte Carlo Results for Structural Estimation (N=1000, 14% Quantile, 10000)

$\sigma = 1.5$		Parameter or Summary Statistic					$\sigma = .25$		True Parameters				
		1.09	1.57	0.25	1.07	1.61			0.30	1.05	1.58	$r\text{-hat}$	0.30
		<i>sd(r-hat)</i>							<i>bias(r-hat)</i>				
		1.99	2.00	2.01	2.05	2.03			2.55	2.57	2.65	1.97	
		<i>pseudo t-statistic</i>											
		0.54	0.81	0.15	0.51	0.78			0.12	0.43	0.59	0.13	
		<i>bias(<math>\sigma\text{-hat}</math>)</i>											
		0.08	0.09	0.09	0.28	0.29	0.29	0.46	0.46	0.46	0.08	0.08	0.08
		<i>sd(<math>\sigma\text{-hat}</math>)</i>											
		2.93	2.89	2.90	3.66	3.61	3.57	3.47	3.43	3.42	<i>pseudo t-statistic</i>		
		<i>bias(<math>\sigma\text{-hat}</math>)</i>											
		0.74	0.69	0.69	0.08	0.08	0.05	0.03	0.03	0.03	<i>pseudo t-statistic</i>		

Parameter or Summary Statistic						True Parameters					
$\sigma = 1.5$				$\sigma = .25$		$\sigma = 1$					
						bias( $\hat{\sigma}$ )					
						sd( $\hat{\sigma}$ )					
						pseudo t-statistic					
7.45	min(mah) :			-7.81	-1.82	-2.82	7.45	8.37	8.15	8.15	7.88
11.87	max(mah) :			12.79	14.36	31.47	12.15	9.43	9.97	9.45	14.91
0.03	0.04	0.08	0.08	0.08	bias( $\hat{\sigma}$ )						
0.29	0.29	0.46	0.46	0.46	sd( $\hat{\sigma}$ )						
3.61	3.67	3.47	3.43	3.43	pseudo t-statistic						



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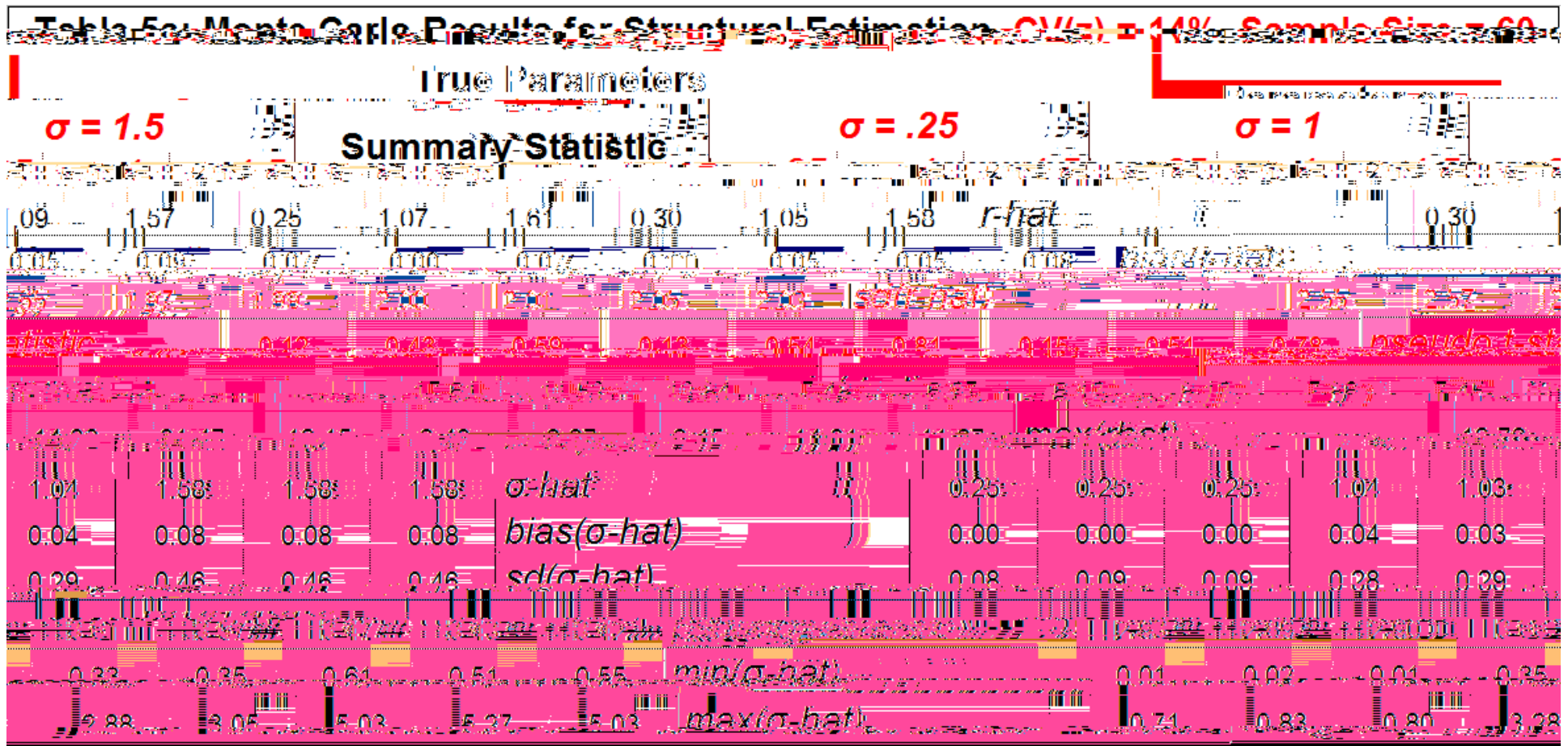
Table 5a: Monte Carlo Results for Structural Estimation,  $CV(\lambda) = 14\%$ , Sample Size = 60

True Parameters										
$\sigma = 1$	0.07	0.00	0.07	0.11	0.05	0.05	0.08	0.30	1.00	1.50
$\text{bias}(\hat{\tau})$	0.07	0.00	0.07	0.11	0.05	0.05	0.08	0.30	1.00	1.50
$\text{pseudo } t\text{-statistic}$	0.12	0.43	0.59	0.13	<b>0.54</b>	0.81	0.15	0.51		
$\text{min}(\hat{\tau})$	17.61	11.52	12.84	7.78	6.37	6.13	8.13	7.86		
$\text{max}(\hat{\tau})$	9.46	14.31	11.87	12.79	14.38	11.47	12.45	9.43		
$\hat{\sigma}$	1.58	1.58	1.58	0.25	0.25	0.25	1.04	1.08		
$\text{set}(\hat{\sigma})$	0.28	0.28	0.48	0.48	0.48	0.08	0.08	0.08		
$\text{pseudo } t\text{-statistic}$	3.61	3.57	3.47	3.43	3.42	2.93	2.89	2.90		
$\text{min}(\hat{\sigma})$	0.00	0.00	0.00	0.00	0.00	0.04	0.02	0.04		
$\text{max}(\hat{\sigma})$	0.45	0.83	0.80	3.38	3.38	3.08	1.92	1.92		









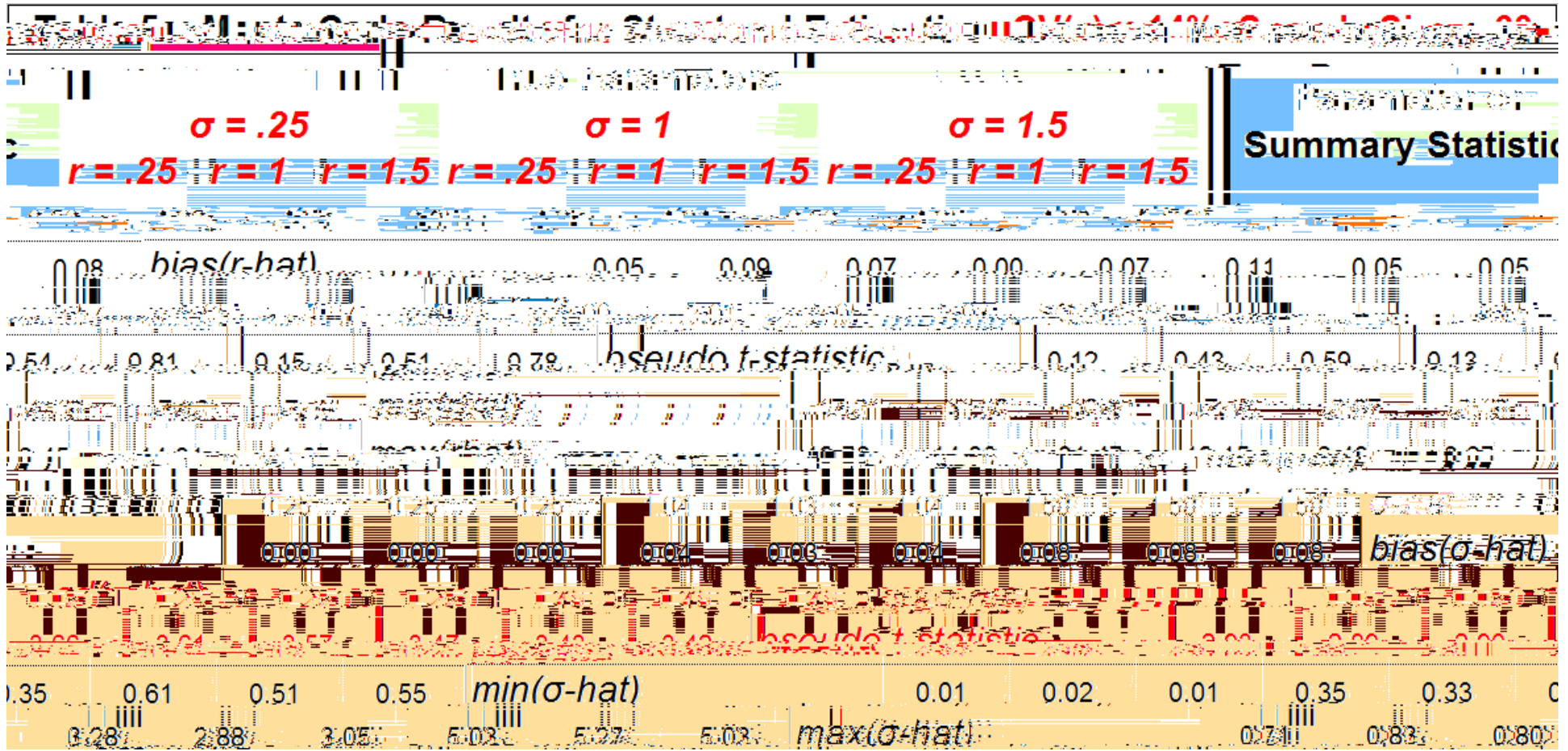
- Punch Line for estimating  $r$  with 60 obs...
  - £ Small bias...
  - £ But unreliable estimates (high variance)

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What About Estimates of ?

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- > Punch line for estimating  $\sigma$  with 60 obs: More reliable estimates, but critically depends on the presence of “good” data on effort
- £ Scaling of  $x_i$  distorts interpretation of
- £ Unreliable if true effort is  $6.570$   $\text{re}Q_6\text{sfrt}391T11$   $\text{A} 1302 1302 0$   $41.22 181T(m)T36.55$ .

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Tullock's  $r$

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Table 4b: Monte Carlo Results for Structural Estimation:  $\tau = 0.25$ ,  $\sigma = 1$ ,  $\rho = 0.5$

Parameter or Statistic	CV(7) = 62%				CV(7) = 1.5%				Variance			
	100	200	400	Sum	100	200	400	Sum				
$\hat{\tau}$	0.25	0.25	0.25	0.25	0.38	0.74	0.21	0.32	0.25	0.25	0.34	0.26
$\sigma\text{-hat}$	0.70	0.99	0.14	0.18	13.25	19.82	7.15	4.27	1.95			
$\text{sd}(\hat{\tau})$	0.03	0.03	0.03	0.03	0.63	0.49	0.07	0.07				
$\text{sd}(\sigma\text{-hat})$	0.36	0.28	0.19	0.24	0.36	0.28	0.19	0.24				
$\text{max}(\hat{\tau})$	59.31	12.15	2.99	15.66	2.17	0.86		361.65	92.71	25.31		
$\sigma\text{-hat}$	1.18	1.04	1.00	1.18	1.04	1.00		1.15	1.04	1.01		
$\text{bias}(\sigma\text{-hat})$	0.16	0.04	0.00	0.16	0.04	0.00		0.15	0.04	0.01		
$\text{sd}(\sigma\text{-hat})$	0.91	0.28	0.19	0.94	0.29	0.19		0.78	0.29	0.19		
$\text{pseudo } t\text{-statistic}$	1.33	0.68	0.04	0.35	0.65	0.04	1.42	2.66	10.09	1.24	2.57	0.92
$\text{max}(\sigma\text{-hat})$	3.04	1.52			26.53	3.04	1.45	35.18	3.28	1.48	53.63	

Table 6b: Monte Carlo Results for Structural Estimation:  $r = 1.5$ ,  $\sigma = 1$

Parameter or	Variation in Effort									
	CV( $\tau$ ) = 1.5%			CV( $\tau$ ) = 14.2%			CV( $\tau$ ) = 68%			
$r\text{-hat}$	1.62	1.51	1.51	1.77	1.92	1.46	1.95	1.61	1.52	2.02
$\text{bias}(r\text{-hat})$	0.12	0.04	0.00	0.27	0.42	0.04	0.45	0.11	0.02	0.52
$\text{sd}(r\text{-hat})$	0.03	0.03	0.03	0.11	0.60	0.03	0.47	0.03	0.03	0.03
$\text{max}(r\text{-hat})$	1.77	1.63	1.60	1.26	1.04	1.00	3.48E+06	1.05	1.01	1.01
$\sigma\text{-hat}$	0.17	0.03	0.00	0.26	0.04	0.00	3.48E+06	0.05	0.01	0.01
$\text{bias}(\sigma\text{-hat})$	0.09	0.03	0.10	0.45	0.29	0.10	3.44E+08	0.32	0.11	0.11
$\text{sd}(\sigma\text{-hat})$	0.04	0.04	0.04	0.14	0.01	0.01	0.01	0.00	0.02	0.02
$\text{max}(\sigma\text{-hat})$	3.79	2.99	1.44	906.14	3.04	1.47	3.44E+10	1.332	1.10	1.10



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# Concluding Remarks

- › Structural estimation of Tullock's  $r$  problematic, unless:
    - £ Have **large sample**, true underlying model has **large  $r$**  and **large variation in  $z$**
  - › Structural estimation of  $r$  requires exceptionally good measures of effort
  - › Suggests utility of developing alternative contest models more amenable to structural estimation
  - › Monte Carlo tests of alternative existing models
  - › Tullock framework still potentially useful for testing predictions via reduced form estimation
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# Concluding Remarks (Continued)

- › Best estimate of  $r$  in antitrust contests brought by the FTC between 1976-2005:  
 $E[r] = 1/4$
  - › Monte Carlo simulations suggest estimate is unbiased, but unreliable (high variance)
-