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Discussion of Durlauf, Navarro and Rivers  
**“Notes on the Econometric Analysis of Crime”**

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The Wharton School, University of Pennsylvania  
CEPR, IZA & NBER

National Academy Workshop, *Understanding Crime Trends*

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# Steve's ~~Advice~~ Complaints

1. Estimation: Start with well-specified choice models
  - Understanding the aggregation of micro-behavior into macro aggregates can guide functional form choices
    - *Ad hoc functional forms*
  - Well-specified models highlight important controls
    - *Policy variables are often correlated (and too often omitted)*
  
2. Interpretation: Policy analysis requires well-specified social welfare functions
  - Requires explicit of counterfactual policy experiments
    - *Counterfactuals rarely well-specified*
  - Statistical significance is not policy significance
    - *Statistical significance is the usually reported*
    - *Cost-benefit relevant magnitudes rarely reported*
  - Probability distributions matter, not just mean effects
    - *Model uncertainty is usually ignored*
    - *Model uncertainty may be large*

# What is Missing?

## My Complaints

0. Poorly-defined policy counterfactuals
1. Incredible instrumental variables
2. Overstatements of precision
3. Publication bias understates model uncertainty

# Complaint #0: Counterfactual Policy Analysis

- What is the relevant counterfactual policy in death penalty analysis?
  1. Abolish the death penalty
    - Test of the “deterrence” hypothesis
  2. Re-allocate all death penalty resources to other criminal justice areas
    - Cost-benefit analysis
  3. Re-allocate “some” proportion of death penalty resources to alternative criminal justice projects and “the rest” to competing state priorities
    - The variation we have.

# Complaint #1: Incredible Instrumental Variables

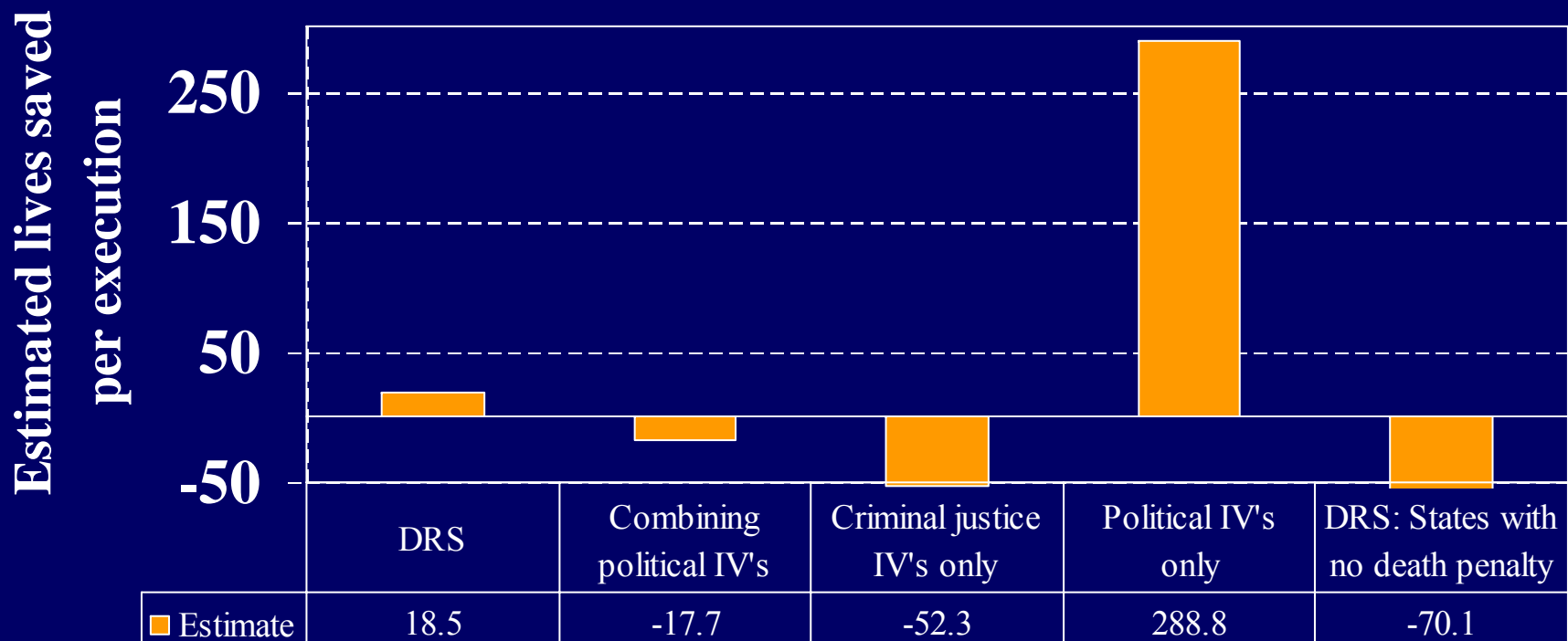
- Identification requires “experiments” in execution policy that do not otherwise affect crime
  - DRS suggest “experiments” in execution from:
    - » State-level police payrolls
    - » State judicial spending
    - » Prison admissions
    - » Partisanship: % of state voting for a Republican Prez (6 variables)
  - Further:
    - » Variables are state aggregates, not per capita
    - » Nominal, rather than real expenditure variables

# An Instrument for All Occasions?

- The instruments used by Dezbakhsh, Rubin and Shepherd have been used in other papers:
  - Lott and Mustard (1997) ⇒ Explain concealed gun laws
  - Rubin and Dezbakhsh (2003) ⇒ Explain concealed gun laws
  - Shepherd (2002a) ⇒ Explain Truth-in-sentencing legislation
  - Shepherd (2002b) ⇒ Explain California’s three strikes laws
  - Shepherd (2004) ⇒ Explain Sentencing guidelines
- In each case, the authors assume:
  - Instrumental variables cause changes in specific deterrence variables
  - But have *no other effects on crime*.
  - ...and hence generate useful “as if” experiments in a particular endogenous variable (and not others!)

# “Overidentification” Tests

## Estimates of Lives Saved per Execution: Alternative “Experiments”



### Alternative IV Specifications

# Assessing the Reduced-Form

	Dependent variable			Net Effect on Homicide Rate <sup>(a)</sup>
	Probability of Arrest	Probability of Death Sentence Given Arrest	Probability of Execution Given Death Sentence	
	(1)	(2)	(3)	
Police Spending	0.03 (0.023)	-0.002*** (0.000)	-0.05*** (0.004)	0.08
Judicial Spending	-0.22*** (0.034)	0.01*** (0.001)	-0.04*** (0.006)	0.58
Prison Admission	0.01*** (0.002)	-0.0001*** (0.000)	0.004*** (0.000)	-0.04
1976 * Republican Vote Share (Ford)	-0.66** (0.311)	0.03 (0.083)	0.49*** (0.053)	0.08
1980 * Republican Vote Share (Reagan I)	0.16 (0.202)	0.004 (0.004)	0.02 (0.036)	-0.45
1984 * Republican Vote Share (Reagan II)	-0.64*** (0.196)	0.04*** (0.004)	0.29*** (0.035)	0.54
1988 * Republican Vote Share (Bush I)	-0.25 (0.216)	0.06*** (0.004)	-0.03 (0.038)	0.41
1992 * Republican Vote Share (Bush II)	-0.04 (0.215)	0.05*** (0.004)	0.14*** (0.039)	-0.45
1996 * Republican Vote Share (Dole)	-0.82*** (0.212)	0.01** (0.004)	0.96*** (0.040)	-0.77
N	48,070	51,143	57,637	
	<b>Second Stage</b>			
Coefficients	-2.27*** (0.50)	-3.62 (14.53)	-2.71*** (0.62)	

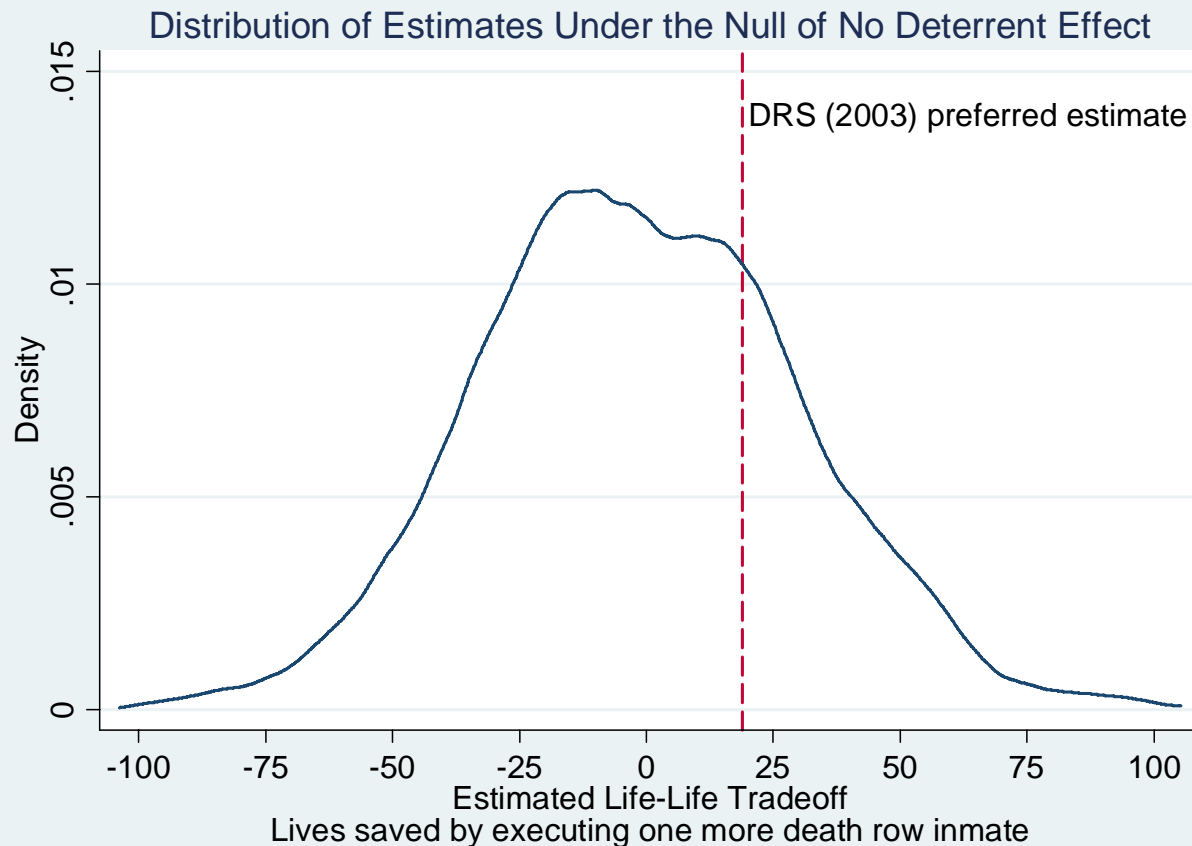


# Rubin's Response

- Donohue and Wolfers claim: These instrumental variables are not possibly “exogenous” shocks to execution policy
  - Fail test of overidentification
- Rubin's response: “Most of our instrumental variables have been used in numerous empirical papers because previous researchers believed (often based on empirical testing) that the instruments were as uncorrelated with crime rates as one was likely to find.”
  - *Economists' Voice*, April 2006 [[Detail](#)]

# Complaint #2: Overstatements of Precision

- DRS treat county-year observations as independent
  - Variation in execution policy (IV's) at state level only
  - Autocorrelation in homicide and execution policy



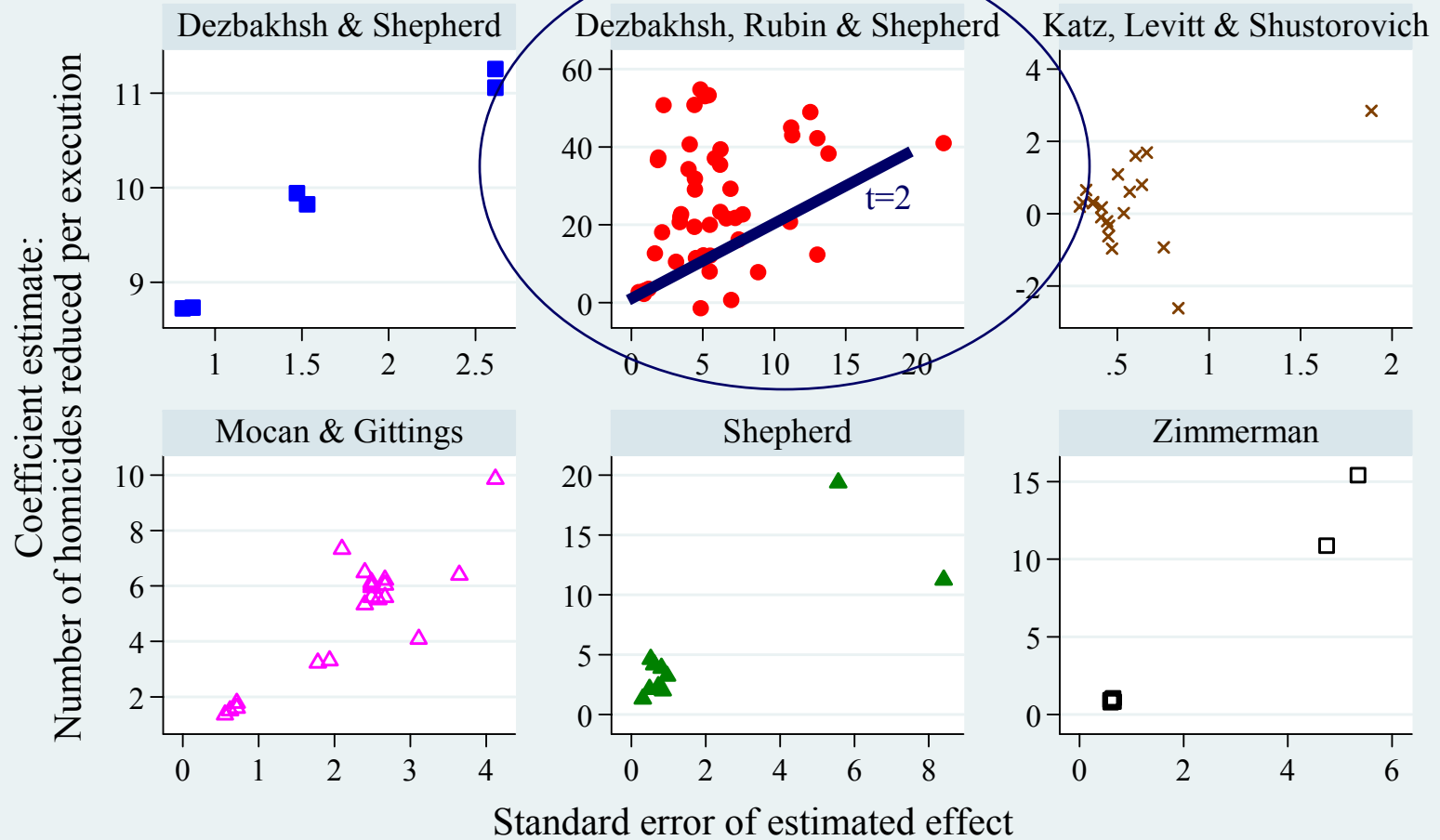
□ *Realistic estimates of parameter uncertainty make Durlauf et al's estimates of model uncertainty less striking*

# Complaint #3: Publication Bias

## Understates Model Uncertainty

### Reporting Bias: Estimated Effects of Executions on Homicide

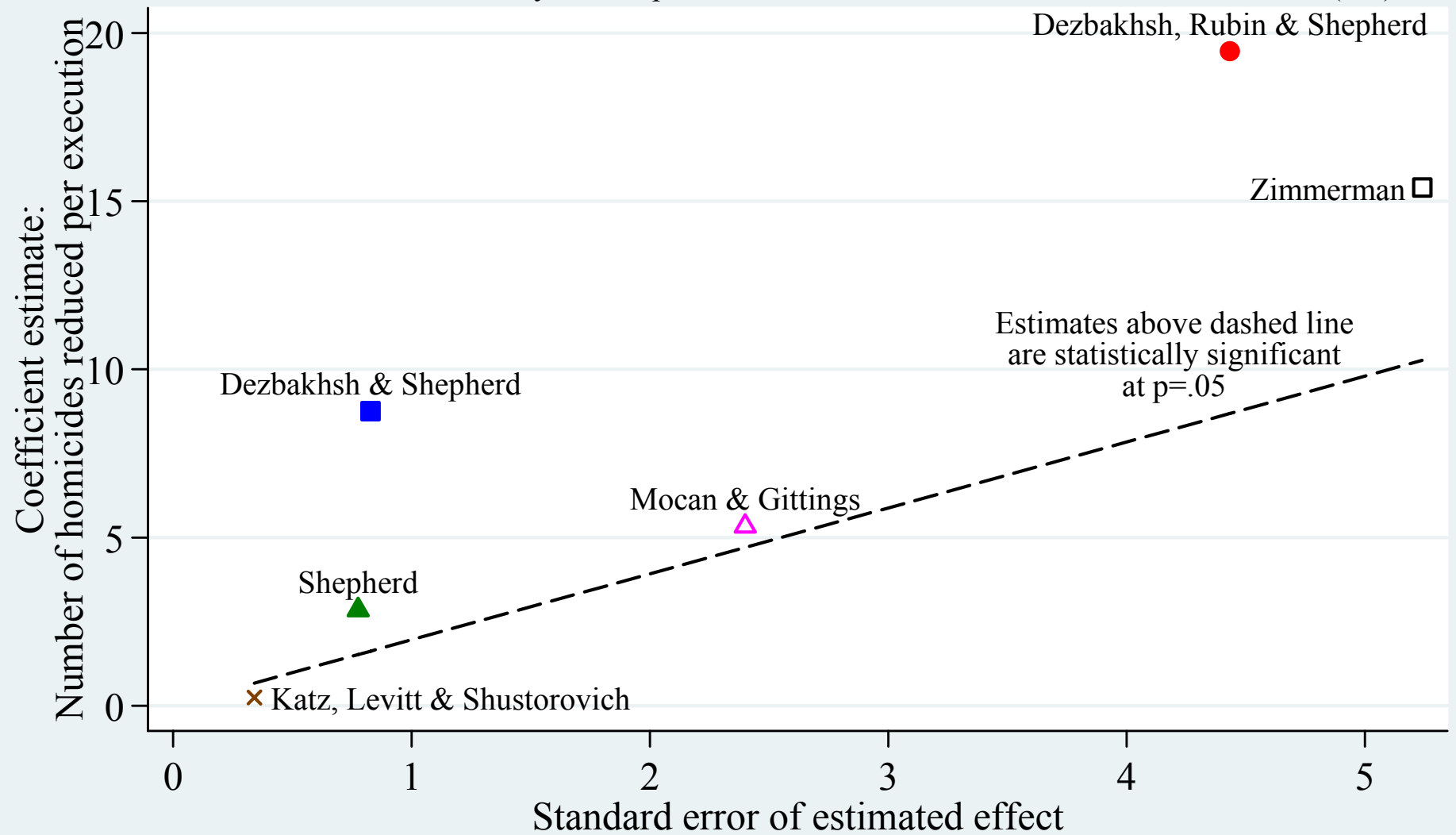
H0: No reporting bias implies that estimated effects should be unrelated to the standard error  
 H1: Results are more likely to be reported if the effect is at least twice the standard error ( $t > 2$ )



Coefficients converted into homicides reduced for the average executing state in 1996

# Reporting Bias: Estimated Effects of Executions on Homicide

H0: No reporting bias implies that estimated effects should be unrelated to the standard error  
H1: Results are more likely to be reported if the effect is at least twice the standard error ( $t > 2$ )



Coefficients converted into homicides reduced for the average executing state in 1996.  
The central estimate from each study is shown.

# What Problem Does Model Averaging Solve?

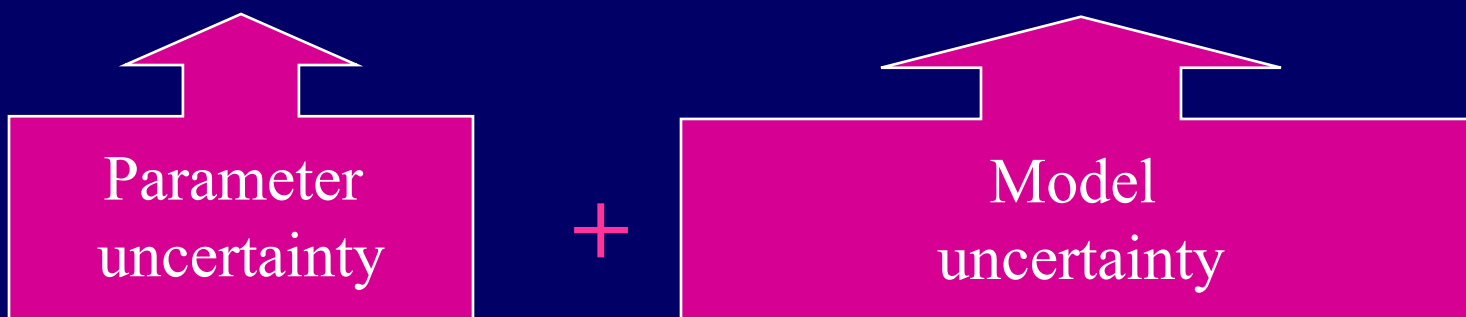
- “Optimal” crime forecast

$$E(\rho_{l,t+1} | D_t, M) = \sum_{m \in M} E(\rho_{l,t+1} | D_t, m) P(m | D_t)$$

- Realistic measure of forecast uncertainty

$$\text{Var}(\rho_{l,t+1} | D_t, M) =$$

$$\sum_{m \in M} \text{Var}(\rho_{l,t+1} | D_t, m) P(m | D_t) + \sum_{m \in M} \left( E(\rho_{l,t+1} | D_t, M) - E(\rho_{l,t+1} | D_t, m) \right)^2 P(m | D_t).$$



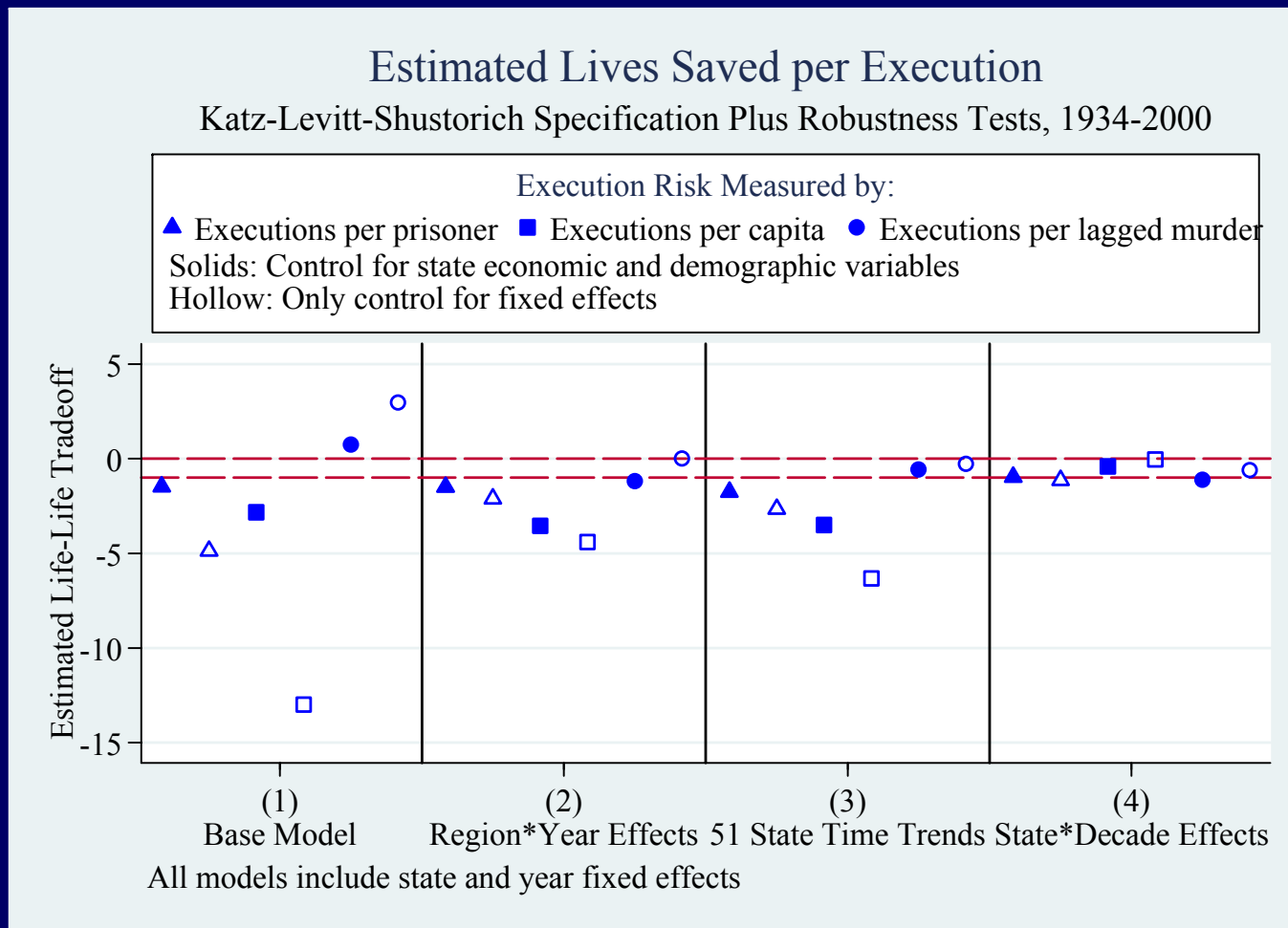
→ We get a reasonable measure of our ignorance

# Which Models to Average?

- Durlauf et al: We should take a posterior weighted average of “coherent models”
- But in reality:
  - Published models are a selected subset
    - » Averaging pro-deterrence studies will yield pro-deterrence averages
  - Some IV studies are not credible
    - » GIGO: “Garbage in, Garbage out”
  - Data cannot speak very precisely
    - » Properly estimated, parameter uncertainty is huge
    - » Model uncertainty also large
      - ◆ But this depends on the range of “coherent” models
- What’s missing?
  - Researcher judgment

# Katz, Levitt, Shustorovich: State Panel Data

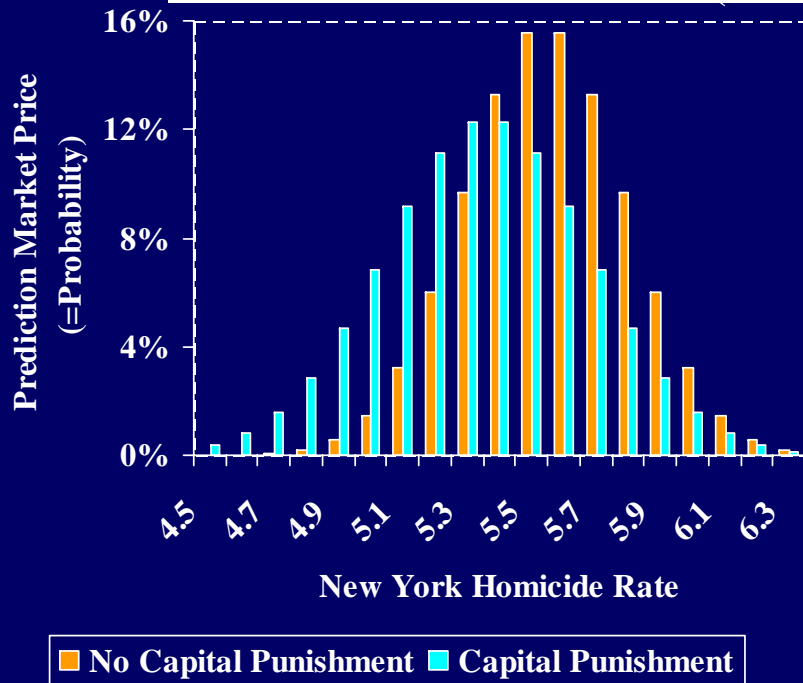
$Homicide\ rate_{s,t} = \beta Execution\ rate_{s,t} + State\ effects_s + Year\ effects_t$   
 + Controls: prisoners per crime, prisoners per capita, prison death rates, real per capita income, %black, %age 02-24, %age 25-44, %urban  
 [+ region\*year effects, state\*time effects, or state\*decade effects]



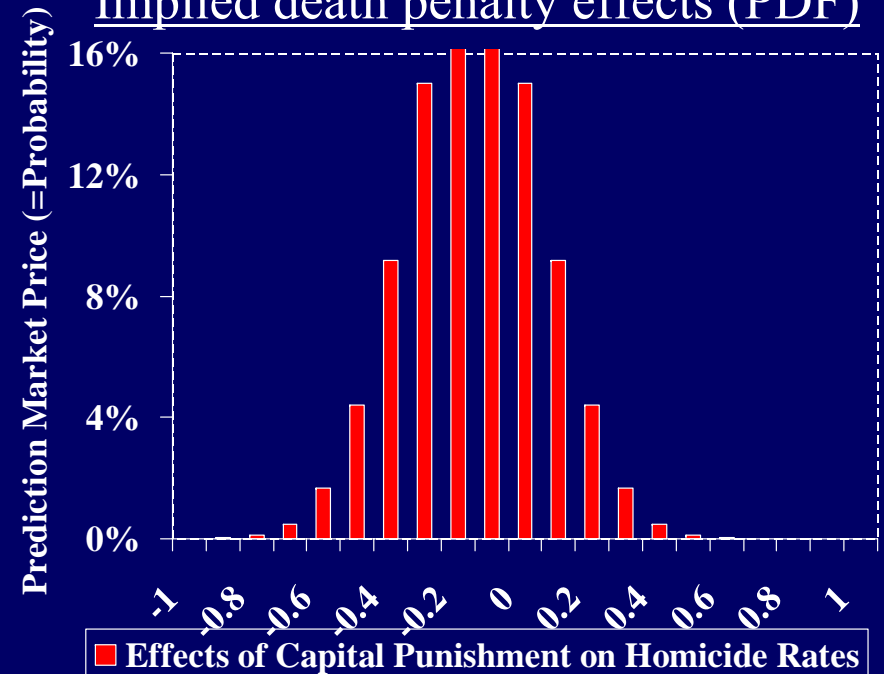
# Model Averaging: A Market-Based Alternative

- Run a prediction market:
  - What will the New York homicide rate be in 2008?
    - » If the Supreme Court allows executions to resume?
    - » If executions are still deemed unconstitutional?

Predicted homicide rates (PDF)



Implied death penalty effects (PDF)





# Model Averaging: A Practical Alternative

Survey 100 clever social scientists.

❖ Sample begins:

1934  1950  1972  1977  1984  2000

❖ Aggregation

National time series  State cross-section  State-year panel  
 State-month panel  county-year panel  OECD country-year panel

❖ Independent variable:

#executions  executions per death sentence<sub>t-6</sub>  executions per murder  
 executions per murder<sub>t-1</sub>  executions per prisoner  executions per capita

❖ Control variables:

Age structure  Racial composition  Incarceration rate  Prison conditions  
 Police  Arrest rate  Death sentences per homicide<sub>t-7</sub>  Non-homicide crime rates  
 State effects  Year effects  State\*year trends  State\*decade effects  Region\*year effects

❖ Estimation:

OLS  WLS  GLS  Median regression  IV  Matching (on what?)

❖ Instruments

Supreme court decisions  State supreme court decisions  Legislative changes  
 Police payrolls  Prison admissions  Judicial spending  % voting Republican  
 Botched executions  Ratio of Black-white homicide rates

❖ Weight you put on this model?

5%  10%  20%  25%  50%  100%