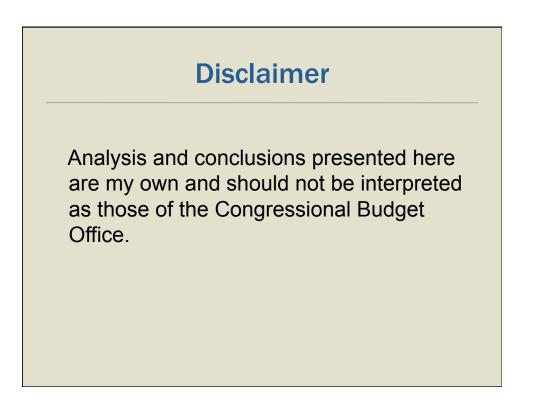
#### Who Will Pay What? Modeling the Distribution of Costs of Climate Change Policies



Kevin Perese Congressional Budget Office Tax Analysis Division

Federation of Tax Administrators Revenue Estimating & Tax Research Conference September 15, 2009 Des Moines, Iowa

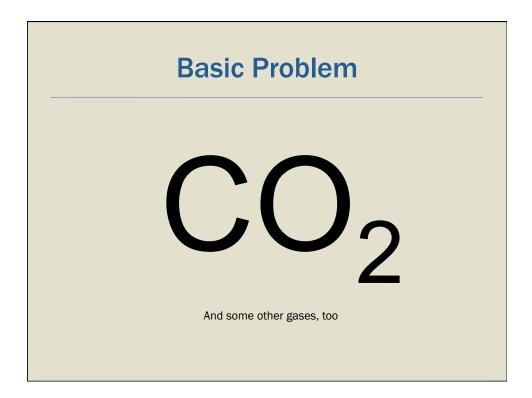


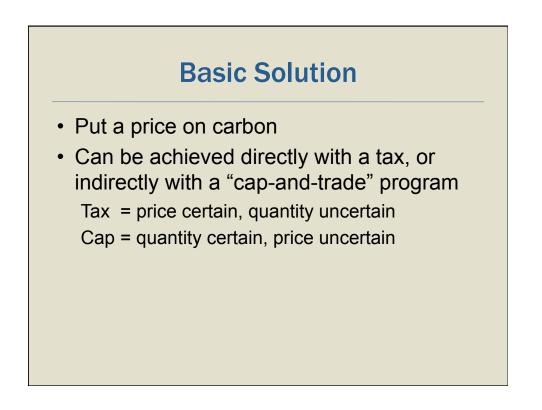
#### **Overview**

- Climate Change Modeling
- Linking Demographic, Tax, and Expenditures Data
- IO Model
- Results

## **Climate Change Modeling**

- Cross-divisional work at CBO
  - Microsimulation of distributional effects small part
  - Estimating allowance price trajectory
  - Budgeting effects
  - International trade effects
  - International & Domestic offsets
  - Transportation & Electricity sector

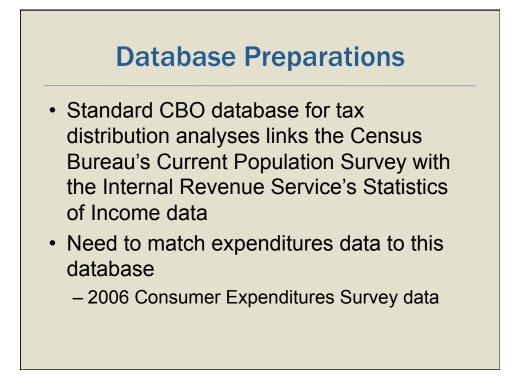




#### Why Microsimulation?

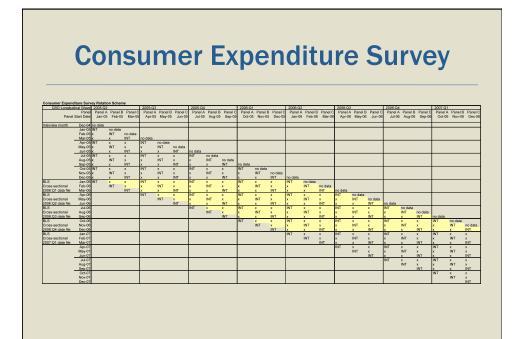
#### **Distributional Analysis**

- Analyze Regressivity/Progressivity of Policies
- Rank by income?
- Rank by expenditures? (permanent income hypothesis)
- Regional analysis



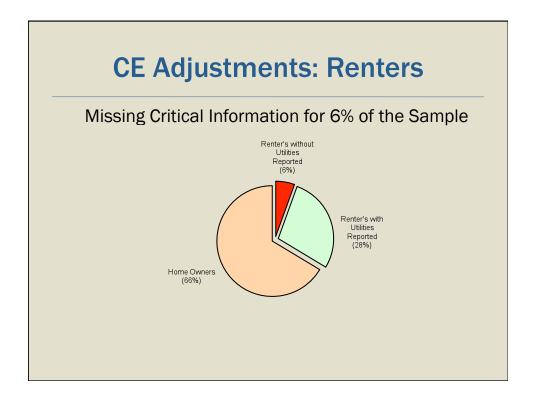
#### **Consumer Expenditure Survey**

- Collects detailed expenditures information on households over 12 month period
- Designed to calculate basket weights in CPI calculation
- Two separate surveys: Interview & Diary
- Released as quarterly cross-sections, but we convert to annual panel files



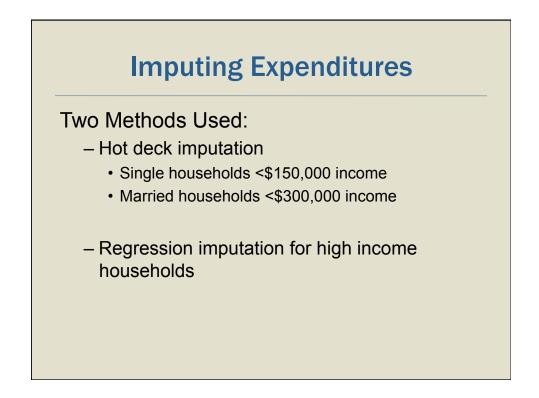
#### **Consumer Expenditure Survey**

		Quintiles								
	All Households	Lowest	Second	Middle	Fourth	Highest				
Dollars										
Electricity	1,303	848	1,104	1,285	1,445	1,831				
Natural Gas	480	273	369	428	559	773				
Fuel Oil	151	82	123	127	177	243				
Total Utilities	1,934	1,203	1,596	1,840	2,181	2,847				
Gasoline & Motor Oil	2,384	1,046	1,768	2,418	2,988	3,696				
Total Energy-Intensive										
Expenditures	4,318	2,249	3,364	4,258	5,169	6,543				
Percent of Income										
Electricity	2.1	8.1	4.0	2.8	2.0	1.2				
Natural Gas	0.8	2.6	1.3	0.9	0.8	0.5				
Fuel Oil	0.2	0.8	0.4	0.3	0.2	0.2				
Total Utilities	3.1	11.4	5.8	4.0	3.0	1.8				
Gasoline & Motor Oil	3.8	9.9	6.4	5.2	4.1	2.3				
Total Energy-Intensive										
Expenditures	6.8	21.4	12.2	9.2	7.1	4.1				



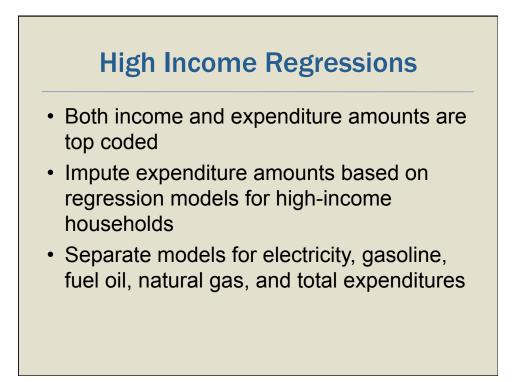
#### **CE Adjustments: Diary Data**

Multiplicative Adjustme	nt Fac	tors
Food Away	*	1.617609
Food at Home	*	0.823904
Alcohol Away	*	1.217858
Alcohol at Home	*	1.634421
Clothing and Shoes	*	1.603027
Furniture	*	1.321044
Additive Adjustment Am	ounts	;
Toiletries	+	\$317.00
Nondurables	+	\$594.47
Prescription Drugs	+	\$130.88
Business Services	+	\$36.79
Tolls	+	\$14.60
Other Education	+	\$70.55



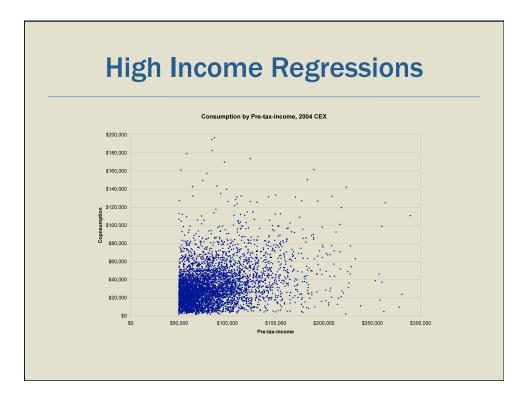
#### Statistical Match SOI/CPS & CE

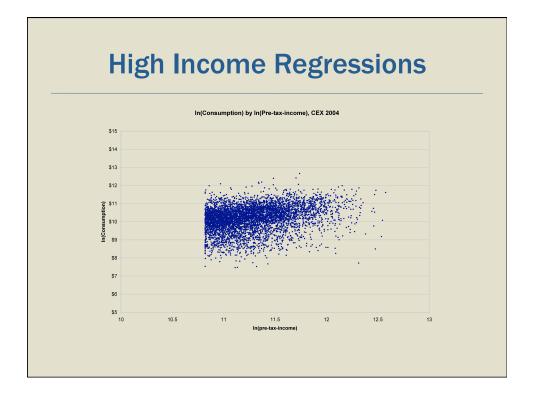
- Hot deck routine with both rigid and flexible matching criteria
  - Fixed: Region
  - Flexible: Age (+/- 1 year increments)
     Income (+/- 2% increments)
     Family Type (+/- 1 child only)
- Use CPS Income as bridge to SOI income
- · Carry over expenditure ratios

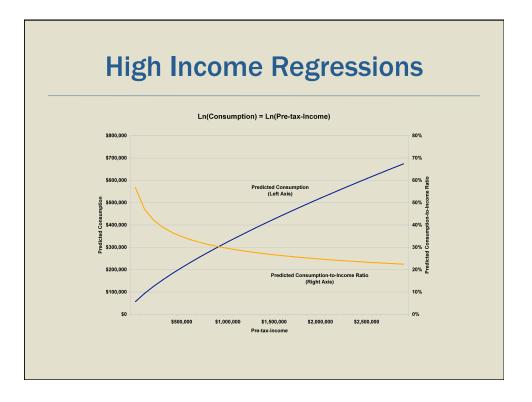


#### **High Income Regressions**

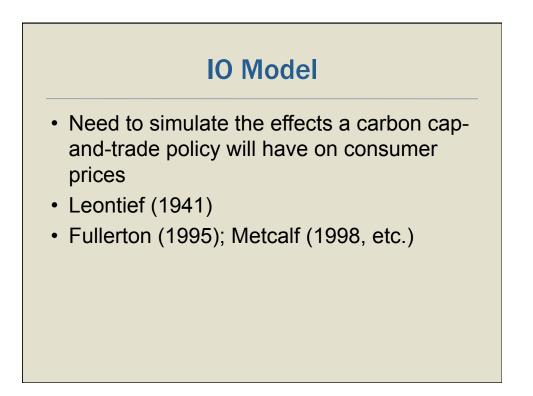
- Estimate electricity, gasoline, fuel oil, natural gas expenditures
- Estimate total expenditures
- Distribute non-carbon intensive expenditures based on observed distribution in high income CE households

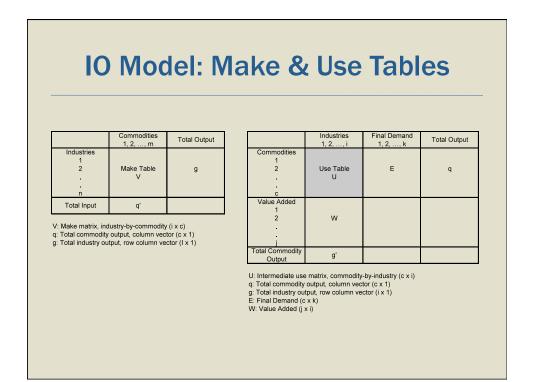






# Now What?





an					$a_{n}p_{n}$				
$a_{\scriptscriptstyle 12}p_{\scriptscriptstyle 1}$ +	+ (				$a_{n_2}p_n$	+	v <sub>2</sub> :		$p_{_2},$ $\vdots$
$a_{_{1n}}p_{_1}$ +	+ c	$a_{2n}p_{2}$	+	 +	$a_{m}p_{m}$	+	$\mathcal{V}_n$	=	$p_{n}$

#### **IO Model: Basic**

$$a_{11}p_{1} + a_{21}p_{2} + \cdots + a_{n1}p_{n} + v_{1} = p_{1},$$

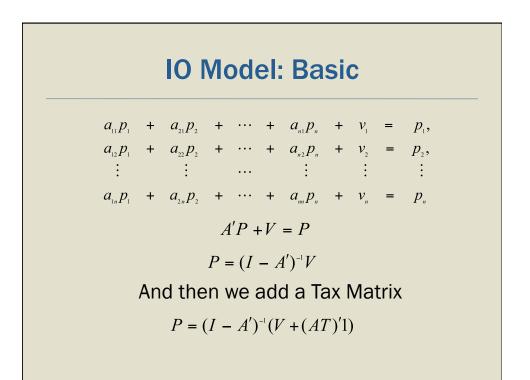
$$a_{12}p_{1} + a_{22}p_{2} + \cdots + a_{n2}p_{n} + v_{2} = p_{2},$$

$$\vdots \qquad \vdots \qquad \vdots \qquad \vdots \qquad \vdots \qquad \vdots$$

$$a_{1n}p_{1} + a_{2n}p_{2} + \cdots + a_{nn}p_{n} + v_{n} = p_{n}$$

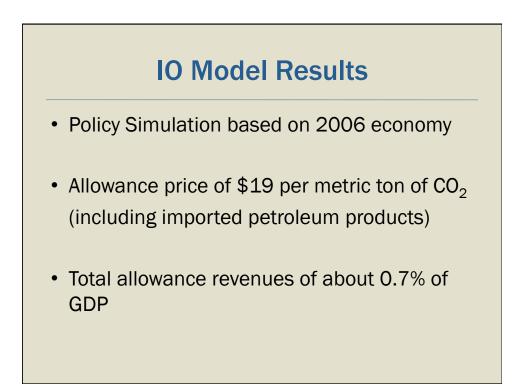
$$A'P + V = P$$

$$P = (I - A')^{-1}V$$



#### IO Model Imports & Non-Combustive Uses

Additional adjustments are made for non-combustive uses of fossil fuels (but I won't bore you further with the equations)



### **IO Model: Price Change Results**

Food	0.5%
Clothing	0.2%
Nondurables	0.4%
Electricity	8.8%
Natural Gas	11.4%
Gasoline	4.2%
All Expenditures	0.7%

