

# OES Time Series

## Alternative designs and estimation methods

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### Research Team

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# Goal of this research

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- Redesign the sample and estimation methods so the design supports both cross-section (or annual) estimates and time series estimates without additional sample units

# A Brief History

- Began with my research
  - ▶ Tested on Alabama
  - ▶ One alternative sample design and one alternative estimation method
  - ▶ Promising Results
  - ▶ Presented to TAC in November 2012
- Major milestones of time series team
  - ▶ Created a simulated OES population
  - ▶ Created/Tested two alternative sample designs
    - Second alternative due to suggestion of TAC
    - Other suggestions made by TAC were incorporated into design
  - ▶ Created/Tested an alternative method for employment and wage estimates

# Current Sample Design

- Frame is created from QCEW records
  - ▶ Over 7 million in-scope establishments
- Sample size = 1.2 million establishments
  - ▶ 400,000 establishments per year (bi-annually)
  - ▶ Large sample size to cover detailed areas and industries
- Allocation, 2 goals:
  - ▶ Minimum allocation (maximize publishable estimates)
  - ▶ Power Neyman allocation (efficiency)
- Selection, stratified PPS
  - ▶ Strata = homogeneous cells (state x area x industry)
  - ▶ PPS = Probability Proportionate to Size (Size = Employment)
  - ▶ **All establishments selected in previous 5 panels are ineligible for selection in current panel**

# Two Alternative Sample Designs

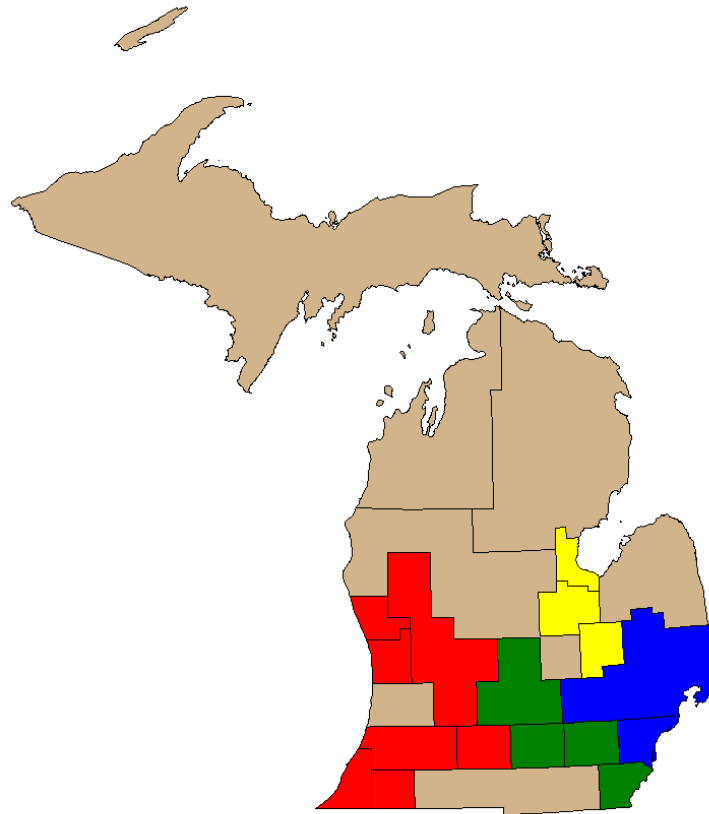
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- Major deviations from current sample design:
  - ▶ An establishment can be sampled in consecutive years
  - ▶ Annual sampling
  - ▶ Any given annual sample represents entire universe
  - ▶ Less detailed strata definitions
- Similarities with current sample design:
  - ▶ Uses similar allocation procedures
  - ▶ Uses probability proportionate to size sampling
  - ▶ Same annual sample size (approx. 400,000 establishments)

# Repeated Cross Section (RCS) Design

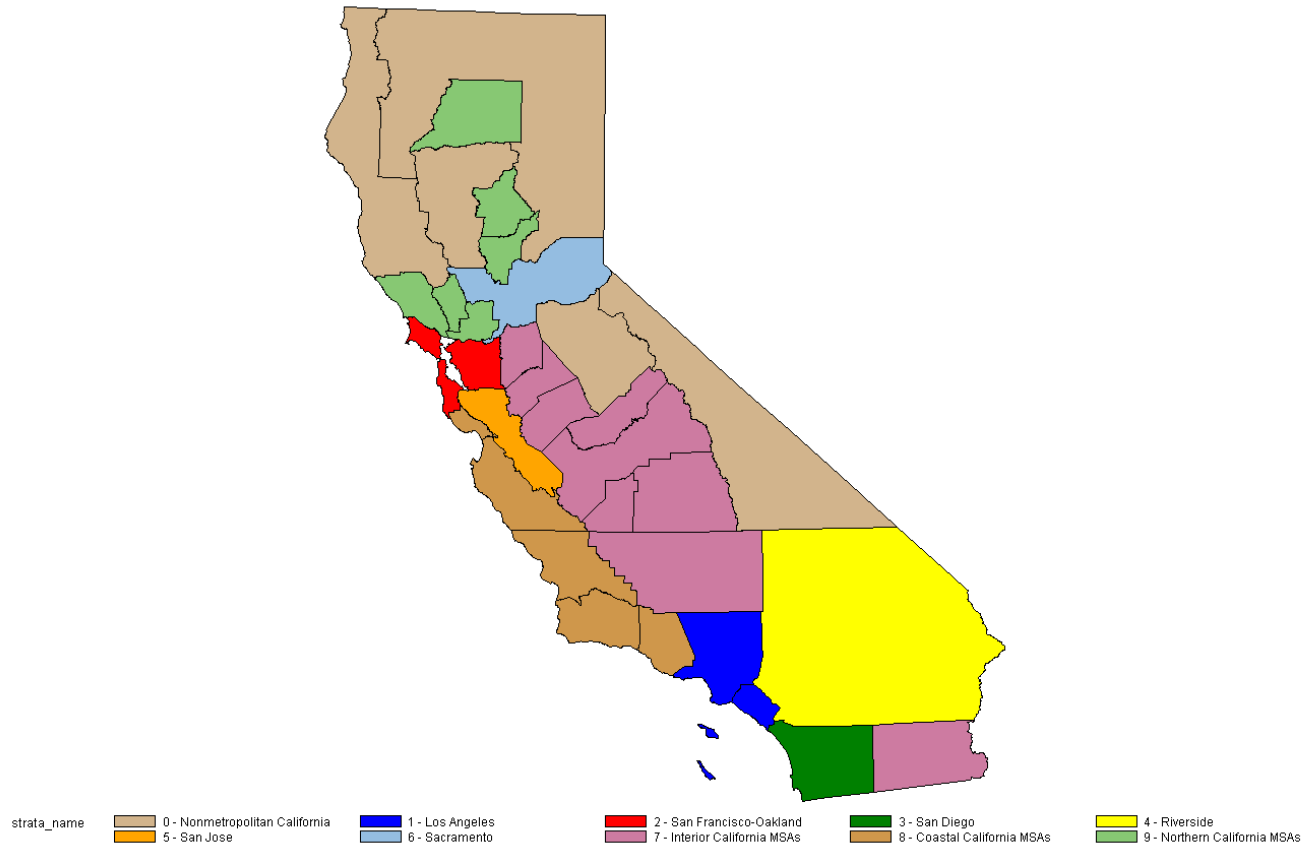
- Stratification: State x Aggregate area x Industry
  - ▶ Aggregate areas: collapse similar areas together within state, based on area size and geographic orientation  
*Example*: MI has 20 MSA/BOS areas → 5 aggregate areas  
CA has 33 MSA/BOS areas → 10 aggregate areas
- Allocation: Minimum + Power Neyman
  - ▶ Minimum Allocation: one per stratum
- Selection: Probability Proportionate to Size (PPS)
  - ▶ **Size = establishment's employment**
- Year-to-Year Overlap: Random

# Michigan Aggregate Areas



strata\_name    0 - Nonmetropolitan MI    1 - Detroit    2 - Western MI MSAs    3 - Southeastern MI MSAs    4 - Northeastern MI MSAs

# California Aggregate Areas





# Two-Thirds Rotating Panel (PAN) Design

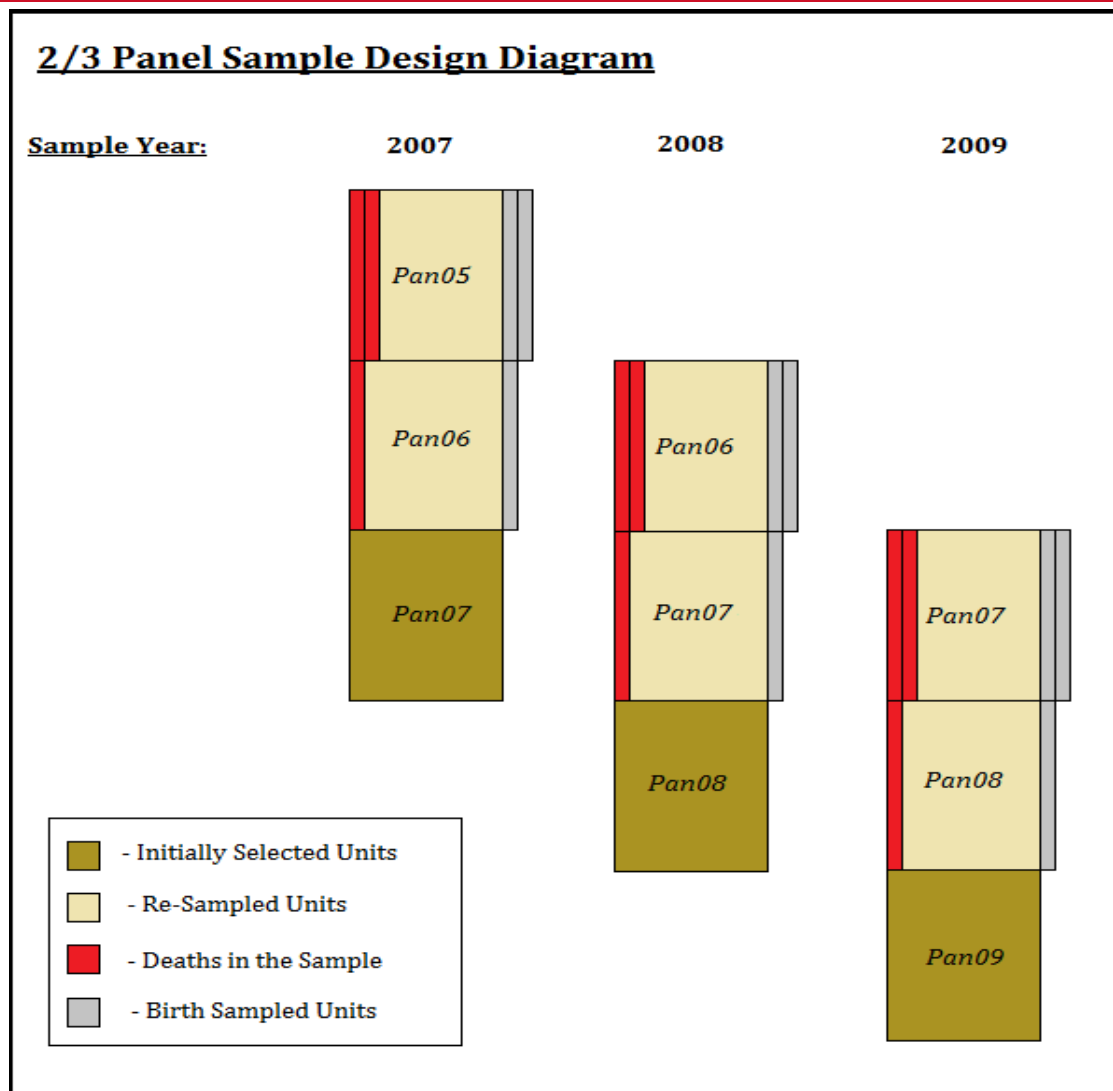
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- Stratification: State x Aggregate area x Industry
- Allocation: Minimum + Power Neyman
  - ▶ Minimum Allocation: three per stratum
- Selection: Probability Proportionate to Size (PPS)
  - ▶ **Size = establishment's employment**
- Year-to-Year Overlap: two-thirds of sample is forced to overlap

# Two-Thirds Rotating Panel (PAN) Design

- Some Details:
  - ▶ Frame is randomly split into three identical sub-frames
  - ▶ Each year only one sub-frame has new allocation and sample
  - ▶ Each year two sub-frames use the same sample from last year
  - ▶ 3 year cycle so that for any given year:
    - One third of the sample is newly allocated/selected
    - One third of the sample was allocated/selected last year
    - One third of the sample was allocated/selected two years ago
  - ▶ Overlapping samples are updated by removing deaths and sampling births

# Two-Thirds Rotating Panel (PAN) Design



# Current Estimation Methods

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- 6 panels of OES data are combined
  - ▶ Employment updated by benchmarking to QCEW
  - ▶ Wages updated by Employer Cost Index (ECI)
- Impute for non-response
- Weight adjustments for atypical data
- Occupation Employment and Mean Wages
  - ▶ Different geography, industry, and ownership detail levels
- Direct estimation (**design-based**)
  - ▶ Sample is weighted up to make estimates
  - ▶ Employment = weighted total
  - ▶ Wages = weighted mean

# Alternative Estimation Method

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- Prediction Theory – use OES respondents to build a model to predict occupational employment/wages for non-responding/non-sampled establishments
  - ▶ End Result: Every establishment on frame has occupational employment/wage data
- Modeled Estimates (*instead of design-based*)
  - ▶ Sample weights used in wage adjustment model
- Occupational Employment and Mean Wages
  - ▶ Different geography, industry, and ownership detail levels
  - ▶ Also can produce change estimates

# Model-Based Estimation Method (MB3): Intuition

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- Goal: predict the staffing pattern and wages of non-sampled units in the target population using sampled establishments (over the previous three years) in the same four or five digit industry and state
- In the estimator, sample units are weighted according to their proximity to the non-sampled unit in four dimensions
  - ▶ Penalty for different six-digit industry
  - ▶ Penalty for different detailed area within the state
  - ▶ Penalty for employment differences
  - ▶ Penalty for sampled in a previous year
- Similar to current employment imputation procedure but without hierarchical structure

# MB3 Example

- Estimation cells
  - ▶ Observationally equivalent establishments
    - Defined by detailed industry - detailed area - employment (continuous)
  - ▶ Some contain lots of establishments, some a single unit
  - ▶ Predictions are identical for all establishments within cell
  - ▶ For computational efficiency purposes only – identical results if processed establishment-by-establishment
- Consider the following estimation cell:
  - ▶ New Single-Family Housing Construction (236115)
  - ▶ Located in Birmingham, AL
  - ▶ With 30 workers
- See table 1, 2.a, and 2.b

**Table 1 -- Characteristics and Weights of the Ten Best Matches**

Match	Area	Strata Area	Sample Year	Reported Employment	Employment Weight	Area and Year Weight	Total Weight	Relative Weight
1	Birmingham	Central Alabama MSAs	2007	49	0.75949	1.00000	0.75949	0.25845
2	Tuscaloosa	Central Alabama MSAs	2007	22	0.84615	0.66667	0.56410	0.19196
3	Gadsden	Central Alabama MSAs	2006	58	0.68182	0.44444	0.30303	0.10312
4	Auburn-Opelika	Southern Alabama MSAs	2007	23	0.86792	0.33333	0.28930	0.09845
5	Mobile	Southern Alabama MSAs	2006	27	0.94737	0.22222	0.21052	0.07164
6	Huntsville	Northern Alabama MSAs	2006	26	0.92857	0.22222	0.20635	0.07022
7	Montgomery	Southern Alabama MSAs	2006	38	0.88235	0.22222	0.19608	0.06672
8	Balance of State Area	Balance of State Areas	2006	39	0.86957	0.22222	0.19324	0.06576
9	Balance of State Area	Balance of State Areas	2005	29	0.98305	0.11111	0.10923	0.03717
10	Huntsville	Northern Alabama MSAs	2005	28	0.96552	0.11111	0.10728	0.03651

Note: Never-responding unit in the 2007 frame is a New Single-Family Housing Construction Company located in Birmingham, AL with 30 workers.

Note: All matches are from the same detailed industry as the never-responding unit.



**Table 2.a -- Reported Employment for Carpenters by Wage Interval and All Other Workers**

Match	Reported Employment	Relative Weight	Carpenters							All Other Workers
			B	C	D	E	F	G	H	
1	49	0.25845		3	5	3	4	1		33
2	22	0.19196		1	5	1	1			14
3	58	0.10312	1	8	3	5				41
4	23	0.09845							4	19
5	27	0.07164	5	7	1	2				12
6	26	0.07022		1	1					24
7	38	0.06672			2		1			35
8	39	0.06576			9	8	1	1		20
9	29	0.03717			2	1	1			25
10	28	0.03651				3	3			22

**Table 2.b -- Employment Shares for Carpenters by Wage Interval and All Other Workers**

Match	Reported Employment	Relative Weight	Carpenters							All Other Workers
			B	C	D	E	F	G	H	
1	49	0.25845		0.06	0.10	0.06	0.08	0.02		0.67
2	22	0.19196		0.05	0.23	0.05	0.05			0.64
3	58	0.10312	0.02	0.14	0.05	0.09				0.71
4	23	0.09845							0.17	0.83
5	27	0.07164	0.19	0.26	0.04	0.07				0.44
6	26	0.07022		0.04	0.04					0.92
7	38	0.06672			0.05		0.03			0.92
8	39	0.06576			0.23	0.21	0.03	0.03		0.51
9	29	0.03717			0.07	0.03	0.03			0.86
10	28	0.03651				0.11	0.11			0.79

# Wage Estimation under MB3

- Estimating interval mean wages
  - ▶ Exploit the fact that the current sample from either RCS or PAN design is representative of the population
  - ▶ Create aggregate occupation and area cells comprised of similarly paid occupations and detailed areas
    - For each occupation, determine what interval the median wage falls into using current sample weighted information (e.g., **nurses'** wages fall in G, postsecondary economics teachers in H, fast food cooks in A)
    - Similarly, for each detailed area, determine what interval the median wage falls into using current sample weighted information (e.g., Boston wages fall in E, Chicago in D, Birmingham in C)
    - Usually around 40 to 50 aggregate occupation x area cells

# Wage Estimation under MB3 (cont'd)

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- Estimating interval mean wages
  - ▶ Assume the wages in each aggregate occupation x area cell follow a unique lognormal distribution
  - ▶ Estimate parameters of each lognormal distribution using maximum likelihood estimation
  - ▶ Graphs
  - ▶ Then directly compute interval means
  - ▶ See Table 3.a

# Middle Wage Occupation Group

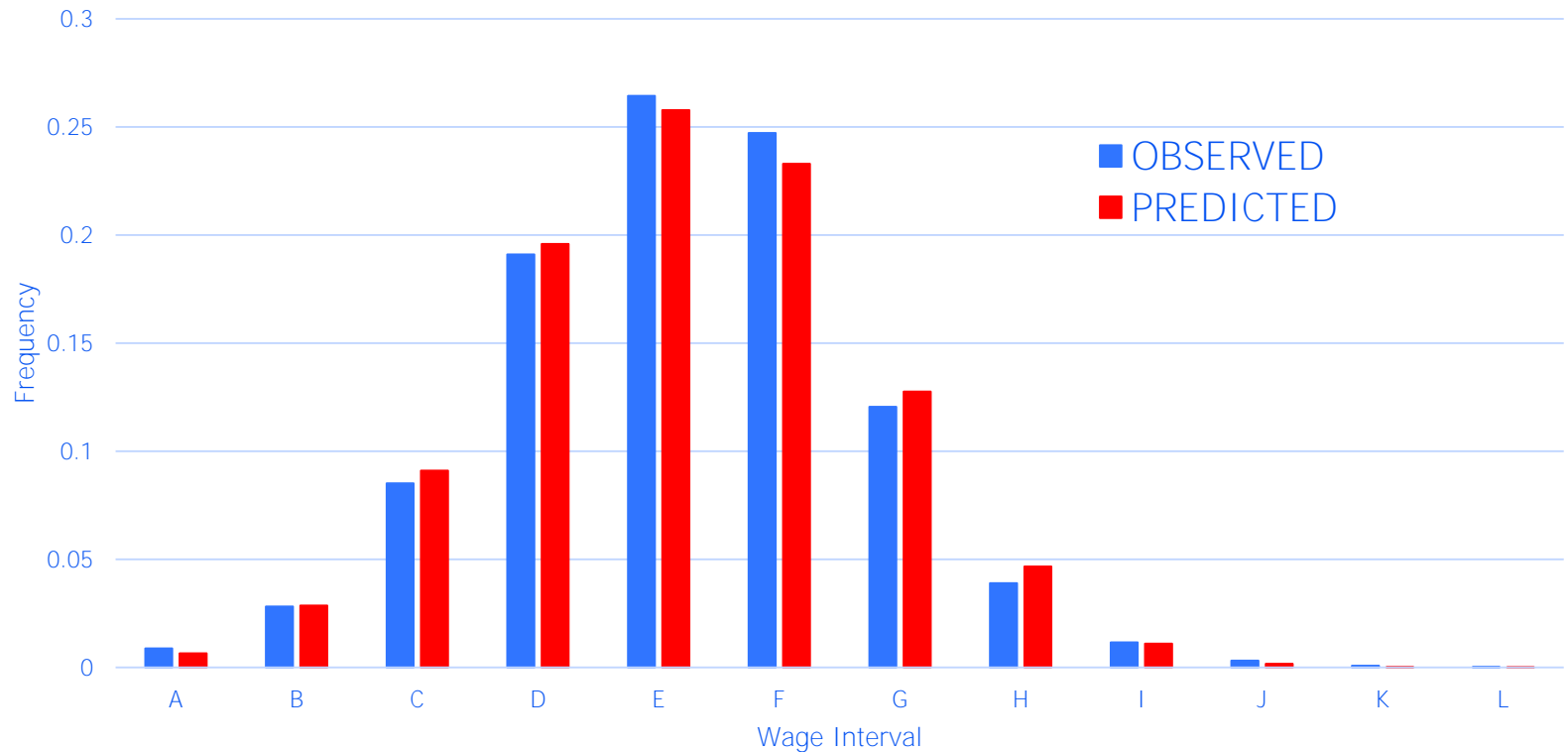


# Middle Wage Occupation Group

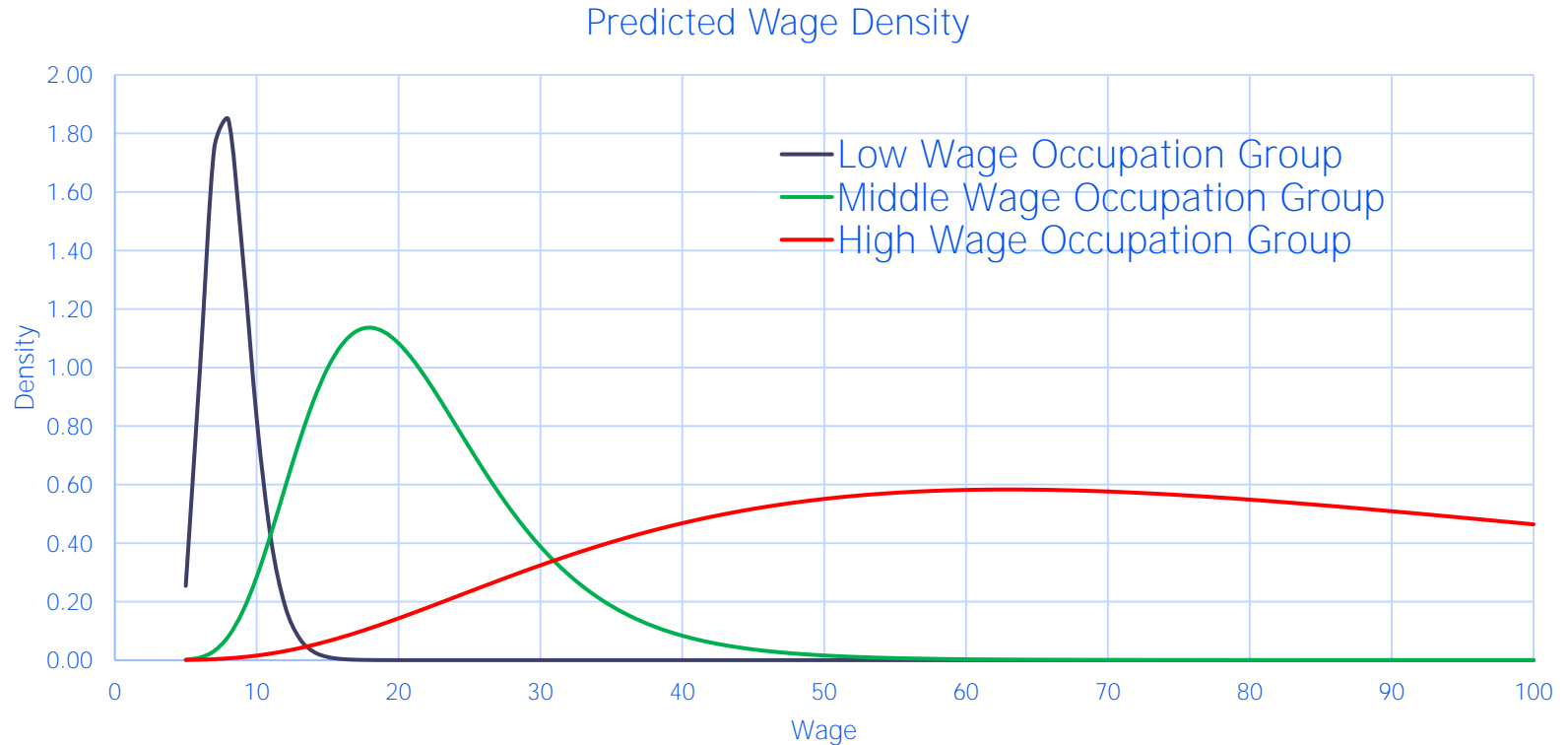


# Middle Wage Occupation Group

Actual vs. Predicted Wage Interval Shares



# Three Predicted Wage Densities



**Table 3.a -- Estimated Interval Mean Wages and Adjustment Factors for Carpenters**

Match	Area	Sample Year	Carpenters							Adjustment factor
			B	C	D	E	F	G	H	
1	Birmingham	2007		10.88	13.67	17.14	21.51	27.05		0.99989
2	Tuscaloosa	2007		10.88	13.67	17.14	21.51			1.05298
3	Gadsden	2006	8.64	10.83	13.60	17.06				1.17902
4	Auburn-Opelika	2007							33.74	1.07270
5	Mobile	2006	8.64	10.83	13.60	17.06				1.13414
6	Huntsville	2006		10.89	13.69					1.06110
7	Montgomery	2006			13.69		21.59			1.09948
8	Balance of State Area	2006			13.60	17.06	21.42	26.96		1.16484
9	Balance of State Area	2005			12.22	15.25	19.09			1.18476
10	Huntsville	2005				15.25	19.09			1.08931

Note: Intervals correspond to sample year. For example, in 2007 interval E includes wages from \$15.25 to \$19.24 per hour, but includes wages from \$13.50 to \$16.99 per hour in 2005.



# Wage Estimation under MB3 (cont'd)

- Under MB3 approach, matches may differ from non-sampled unit in four dimensions:
  - ▶ Area
  - ▶ Industry
  - ▶ Year
  - ▶ Employment
- Need to adjust wages of matches into current, local dollars
  - ▶ Standard log wage regression on current data
  - ▶ Main effects for occupation, area, industry by strata area, and employer size
  - ▶ Coefficients vary across time

# Wage Estimation under MB3 (cont'd)

- Adjustment factor
  - ▶ Ratio of predicted wage for an occupation in non-sampled establishment to predicted wage for an occupation in the match
  - ▶ For every occupation reported by a match there is a unique adjustment factor
  - ▶ See table 3.a
- Interval shifts
  - ▶ Adjusted wages in current dollars
  - ▶ May need to shift wage interval employment if adjusted wage falls into a different interval
  - ▶ See table 3.b and 4.a

**Table 3.a -- Estimated Interval Mean Wages and Adjustment Factors for Carpenters**

Match	Area	Sample Year	Carpenters							Adjustment factor
			B	C	D	E	F	G	H	
1	Birmingham	2007		10.88	13.67	17.14	21.51	27.05		0.99989
2	Tuscaloosa	2007		10.88	13.67	17.14	21.51			1.05298
3	Gadsden	2006	8.64	10.83	13.60	17.06				1.17902
4	Auburn-Opelika	2007							33.74	1.07270
5	Mobile	2006	8.64	10.83	13.60	17.06				1.13414
6	Huntsville	2006		10.89	13.69					1.06110
7	Montgomery	2006			13.69		21.59			1.09948
8	Balance of State Area	2006			13.60	17.06	21.42	26.96		1.16484
9	Balance of State Area	2005			12.22	15.25	19.09			1.18476
10	Huntsville	2005				15.25	19.09			1.08931

Note: Intervals correspond to sample year. For example, in 2007 interval E includes wages from \$15.25 to \$19.24 per hour, but includes wages from \$13.50 to \$16.99 per hour in 2005.

**Table 3.b -- Adjusted Interval Mean Wages for Carpenters**

Match	Area	Sample Year	Carpenters							Interval Shift
			B (\$7.50 to \$9.49)	C (\$9.50 to \$11.99)	D (\$12.00 to \$15.24)	E (\$15.25 to \$19.24)	F (\$19.25 to \$24.49)	G (\$24.50 to \$30.99)	H (\$31.00 to \$39.24)	
1	Birmingham	2007		10.88	13.67	17.14	21.51	27.05		
2	Tuscaloosa	2007		11.46	14.39	18.05	22.65			
3	Gadsden	2006		10.19	12.77	16.03	20.11			Yes
4	Auburn-Opelika	2007							36.19	
5	Mobile	2006		9.80	12.28	15.42	19.35			Yes
6	Huntsville	2006		11.56	14.53					
7	Montgomery	2006			15.05		23.74			
8	Balance of State Area	2006				15.84	19.87	24.95	31.40	Yes
9	Balance of State Area	2005			14.48	18.07	22.62			
10	Huntsville	2005				16.61	20.79			

Note: Intervals now correspond to the current year. The interval shift column indicates whether wages (and subsequently employment) moved from one interval to another.

**Table 4.a -- Adjusted Employment Shares for Carpenters by Wage Interval and All Other Workers**

Match	Reported Employment	Relative Weight	Carpenters							All Other Workers
			B (\$7.50 to \$9.49)	C (\$9.50 to \$11.99)	D (\$12.00 to \$15.24)	E (\$15.25 to \$19.24)	F (\$19.25 to \$24.49)	G (\$24.50 to \$30.99)	H (\$31.00 to \$39.24)	
1	49	0.25845		0.06	0.10	0.06	0.08	0.02		0.67
2	22	0.19196		0.05	0.23	0.05	0.05			0.64
3	58	0.10312		0.02	0.14	0.05	0.09			0.71
4	23	0.09845							0.17	0.83
5	27	0.07164		0.19	0.26	0.04	0.07			0.44
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7	38	0.06672			0.05		0.03			0.92
8	39	0.06576				0.23	0.21	0.03	0.03	0.51
9	29	0.03717			0.07	0.03	0.03			0.86
10	28	0.03651				0.11	0.11			0.79

# Wage Estimation under MB3 (cont'd)

- Putting it all together
  - ▶ For each match we have occupation - (current) wage interval employment shares (table 4.a) and mean wages (table 3.b)
  - ▶ Predicted employment for an occupation-interval is weighted sum over matches
    - Sum of Employment Share x weight x employment of non-sampled unit
  - ▶ Predicted total wage for an occupation-interval is weighted sum over matches
    - Sum of Employment Share x mean wage x weight x employment of non-sampled unit
  - ▶ See table 4.b

**Table 4.b -- Weighted Employment Levels for Carpenters by Wage Interval and All Other Workers**

Match	Reported Employment	Relative Weight	Carpenters							All Other Workers	Total
			B (\$7.50 to \$9.49)	C (\$9.50 to \$11.99)	D (\$12.00 to \$15.24)	E (\$15.25 to \$19.24)	F (\$19.25 to \$24.49)	G (\$24.50 to \$30.99)	H (\$31.00 to \$39.24)		
1	49	0.25845		0.47	0.79	0.47	0.63	0.16		5.22	7.75
2	22	0.19196		0.26	1.31	0.26	0.26			3.66	5.76
3	58	0.10312		0.05	0.43	0.16	0.27			2.19	3.09
4	23	0.09845							0.51	2.44	2.95
5	27	0.07164		0.40	0.56	0.08	0.16			0.96	2.15
6	26	0.07022		0.08	0.08					1.94	2.11
7	38	0.06672			0.11		0.05			1.84	2.00
8	39	0.06576				0.46	0.40	0.05	0.05	1.01	1.97
9	29	0.03717			0.08	0.04	0.04			0.96	1.12
10	28	0.03651				0.12	0.12			0.86	1.10
<b>Predicted Employment</b>				1.27	3.35	1.59	1.93	0.21	0.56	21.09	30.00
<b>Predicted Mean Wages</b>				10.67	13.69	16.70	20.99	26.54	35.76		

# Testing the Alternative Methods

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- Created a fictional population where occupational employment and wages were known for all establishments
  - ▶ Covered 18 states
  - ▶ Include small / medium / large (within limit) states across all census regions
  - ▶ Responders use their data directly
  - ▶ Non-responders use mixing algorithm to borrow from across both time and space
  - ▶ Covers 2004 to 2009
- Created population = truth for testing



# Testing the Alternative Methods

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- Used a Simulation Study
- Selected 100 samples using current, RCS, and PAN sample designs
- Create estimates using current, and proposed model methods
- Did this 6 times – 2004 to 2009
- Calculate statistics showing how well each method does at estimating yearly and change estimates

# Summary of Results

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- Let's look at tables of results (separate document)
- The alternative methods systematically outperform the current OES methods
- Current methods performs the best when occupational estimates are stable
- The model based estimates using the RCS design have the best overall results
- The model based estimates using the PAN design have comparable (and sometime better) results than the model based estimates using the RCS design for estimates of change

# Next Steps

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- Research phase is coming to an end
  - ▶ Methods for measuring reliability still need to be developed
  - ▶ Possible testing of alternative allocations
- Decision phase is beginning
  - ▶ Should we pursue these methods?
  - ▶ How would we implement these methods?
    - What tools are needed for implementation?
    - What is the most efficient way to implement these methods?
    - How will we determine the aggregate areas to sample by?
    - Etc.

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Thanks!