

Modeling Uninsurance Estimates at the County Level

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BACKGROUND

The Patient Protection and Affordable Care Act (ACA) marketplace and Medicaid expansions were fully implemented in 2014

- Many of the provisions implemented by states
- Health information at low levels of geography can lead to action at local levels to improve the health of a community

Sub-state data are needed for implementation and evaluation of the ACA

- Inform outreach and enrollment efforts
- Monitor and evaluate the impact and effectiveness of the law

DATA

Individual state-based surveys

- Pros: Tailored questions focused on particular situations and programs within a state; timely; often official source of state uninsurance estimates
- Cons: Often lack sufficient sample to produce reliable estimates at the county level

American Community Survey (ACS)

- Pros: Estimates available for all counties from 5-year products
- Cons: Pooling 2014 with several years prior to ACA implementation is difficult to interpret

Small Area Health Insurance Estimates (SAHIE) Program

- Pros: Single year estimates available for all counties
- Cons: Only provides a limited set of uninsurance estimates with a delayed data release

RESEARCH OBJECTIVE

Produce uninsurance rates for all counties in Oklahoma for the non-elderly population. The objective was to use the state-based survey, the Oklahoma Health Care Insurance and Access Survey (OHIS). We evaluated multiple options based on estimates of uncertainty.

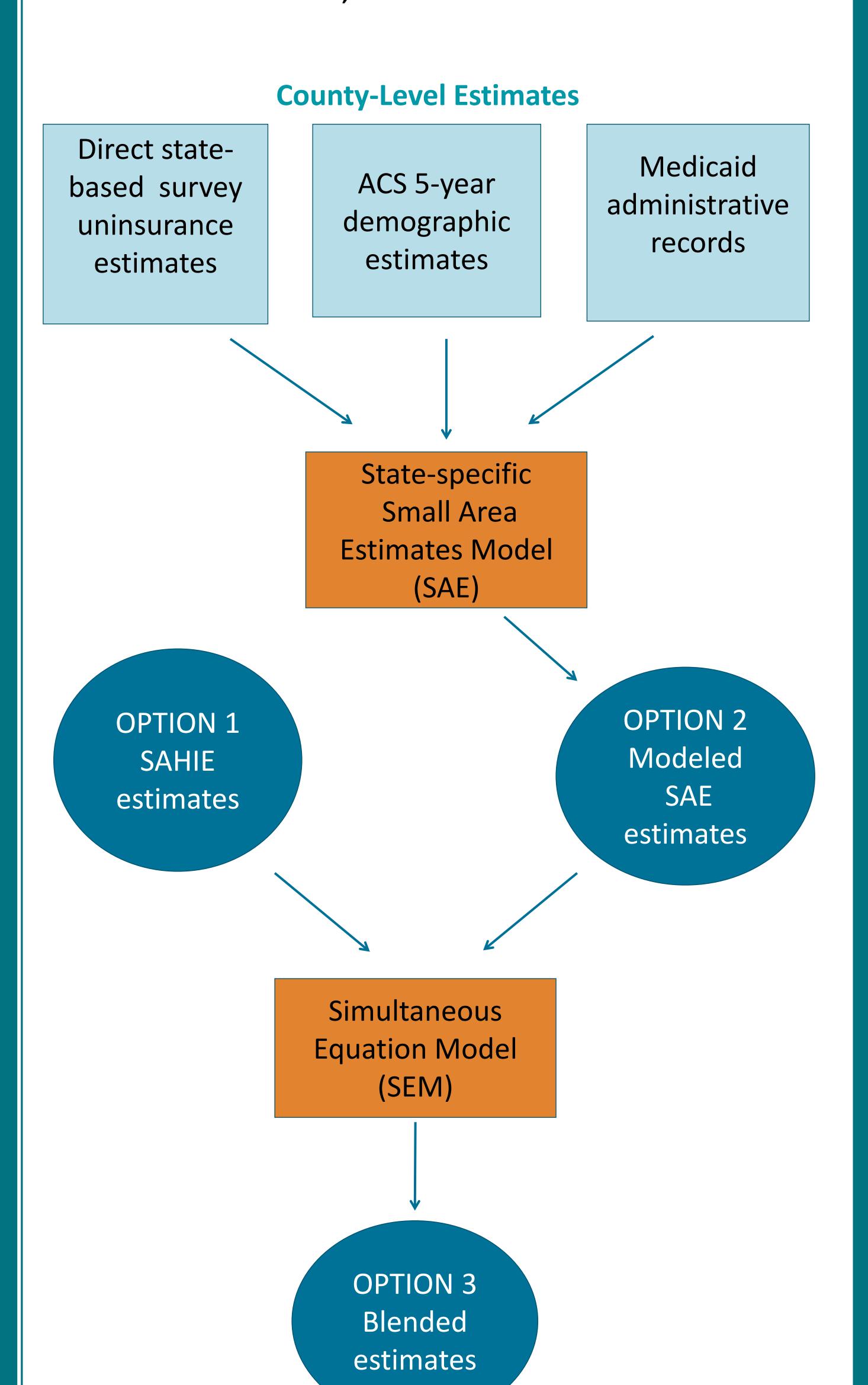
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METHODOLOGY

We constructed the models in steps so multiple options could be considered

Most current data available at the time of the analysis

 2013 state OHIS survey, 2007-2011 ACS, 2012 Medicaid administrative records, and 2011 SAHIE



RESULTS

Comparison of descriptive statistics across all counties in Oklahoma

| | Direct Estimates (%) | SAHIE (Option 1) (%) | SAE Model (Option 2) (%) | SEM Model (Option 3) (%) |
|-------------------------------|----------------------------|----------------------------|--------------------------------|--------------------------------|
| Estimates | | | | |
| High | 61.4 | 28.6 | 37.3 | 27.9 |
| Low | 0.0 | 15.7 | 14.3 | 15.8 |
| Average | 22.5 | 22.9 | 20.9 | 22.6 |
| Coefficient of Variation (CV) | | | | |
| High | 103.4 | 6.4 | 31.1 | 5.9 |
| Low | 9.3 | 2.4 | 7.6 | 2.1 |
| Average | 39.7 | 4.7 | 14.7 | 4.5 |
| CV > 30% | 62.5 | 0.0 | 1.3 | 0.0 |

Estimates from Option 2 are appealing as they rely solely on uninsurance as measured by the state-based survey, but the predicted county level estimates have high variance

Estimates from Option 3 use multiple sources of the outcome variable, taking advantage of the greater precision of the SAHIE estimates while still using the state-based survey

Option 3 was selected by Oklahoma

CONCLUSIONS

Reliable sub-state data is needed to monitor and evaluate the ACA. As modeling estimates becomes more common it's important to understand the strengths and weaknesses of the methods

Having SAHIE, essentially a benchmark, was unique and demonstrated the importance of providing measures of uncertainty and transparency in documentation

Given the precision of SAHIE, modeling uninsurance using state-based survey data may not yield much gain, but the approach can be used for other outcome variables