

### Modeling Evidence-Based Interventions (EBI) Impact Workgroup Update

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### **Modeling EBI Impact Workgroup Objectives**

- Inform cancer screening-focused EBI implementation planning, practicelevel change, and policies at the state and national levels
- Use models to simulate and compare the health and economic impacts of alternate "what if" scenarios on:
  - Cancer screening and outcomes in a given region over time
  - The percent of sub-populations up-to-date with recommended screening, as well as changes in cancer incidence, cancer stage at diagnosis, cancer deaths and/or life-years lost due to cancer
  - Comparative costs and cost-effectiveness of cancer screeningfocused interventions
- Integrate best available evidence to evaluate uncertainty







Stephanie Wheeler, UNC

### **Our Team**



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Sarah Drier, UNC



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Prevention Research Centers

(Not Shown): Meghan O'Leary, UNC



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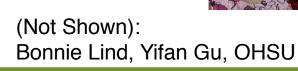


Jackie Shannon, OHSU



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Gloria Coronado, **KPCHR, UW** 



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### **Current Research Foci**

- To estimate the impact of health insurance expansion in North Carolina on colorectal cancer (CRC) screening and outcomes (simulation)
- 2. To estimate the impact of health insurance expansion in Oregon on CRC screening and outcomes (simulation)
- **3**. To estimate the impact of CRC screening EBIs in Oregon on CRC outcomes (simulation)
- 4. To understand variation in CRC screening within Oregon's Medicaid Coordinated Care Organizations (CCOs)
- 5. To understand how Oregon's CCOs have increased CRC screening and through what mechanisms

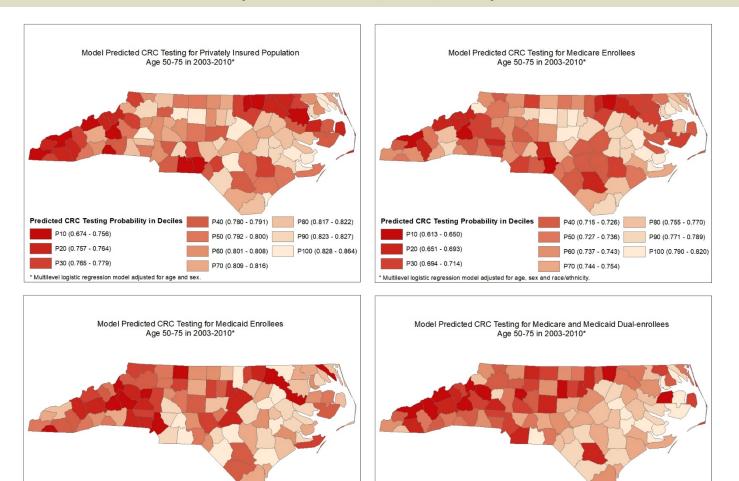


### **Research Activities: May 2017-Present**

- Re-parameterized simulation model with updated input parameter estimates from claims data, BRFSS and other sources
- Troubleshooted preliminary outputs and re-evaluated modeling assumptions
- Consulted with the CISNET MISCAN-Colon modeling group about modeling approach and natural history parameters
- Analyzed Oregon claims data to isolate the effects of health insurance expansion
- Supported Oregon analyses of CCO patterns of care related to CRC screening initiatives

### **Colorectal cancer screening varies greatly by** county in NC insured populations

(Wheeler et al, H&P, 2014)



\* Multilevel logistic regression model adjusted for age, sex and race/ethnicity. Prevention Research Centers

Predicted CRC Tesing Probability in Deciles

P10 (0.554 - 0.591)

P20 (0.592 - 0.600)

P30 (0 601 - 0 609)

P40 (0.610 - 0.618)

P50 (0.619 - 0.625)

60 (0.626 - 0.633)

P70 (0.634 - 0.642)

P80 (0.643 - 0.650)

P90 (0.651 - 0.664)

P100 (0.665 - 0.689)

Predicted CRC Testing Probability in Deciles

\* Multilevel logistic regression model adjusted for age, sex and race/ethnicity.

P10 (0.534 - 0.581)

P20 (0.582 - 0.596)

P30 (0.597 - 0.615)

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P80 (0.666 - 0.679)

P90 (0.680 - 0.704)

P100 (0.705 - 0.766)

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P40 (0.616 - 0.629)

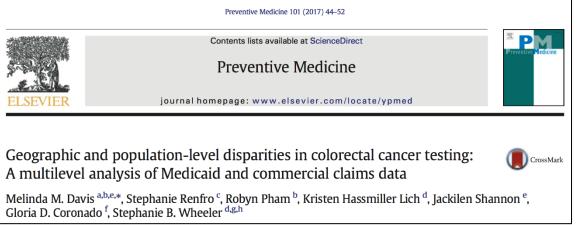
P50 (0.630 - 0.639)

P60 (0.640 - 0.654)

P70 (0.655 - 0.665)

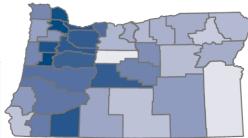
### North Carolina is not unique in its regional variation in CRC screening (psst: Oregon!)

(Davis et al, Prev Med, 2017)

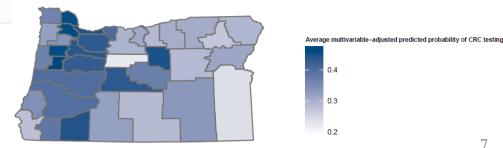


Overall

Medicaid



Commercial



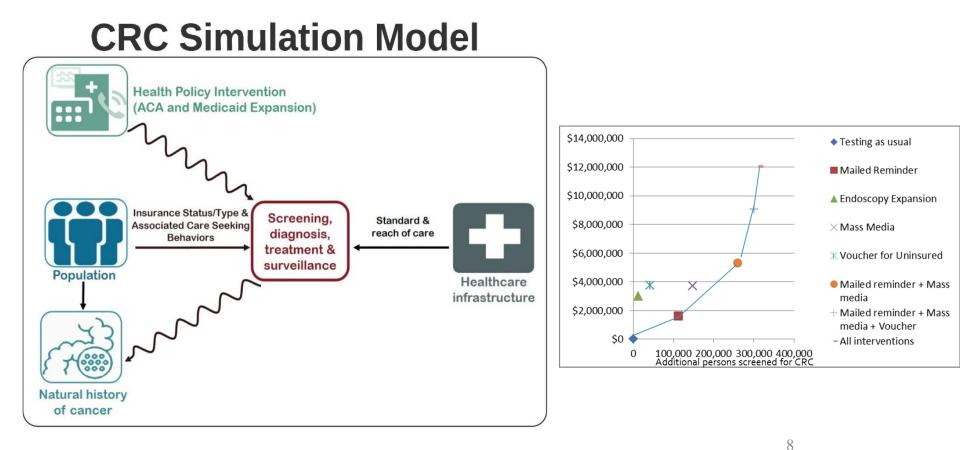


#### Highlights

- Despite insurance, 58% had not received colorectal cancer (CRC) testing.
- CRC testing varied from 22.4% to 46.8% across Oregon's 36 counties.
- Individual, community, and health system-level factors impacted CRC testing.
- · Counties with higher socioeconomic deprivation displayed lower CRC testing.
- · Work to increase CRC testing in targeted counties and populations is needed.



### Mathematical simulations can help us understand which strategies are expected to be most costeffective, where and for whom (Hassmiller Lich et al, PCD, 2017)

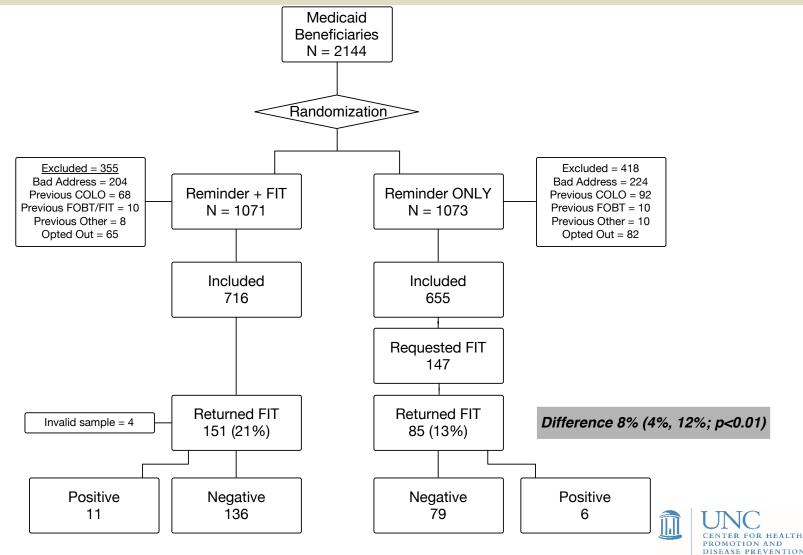






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# CRC FIT-based outreach partnerships with state Medicaid coordinated care organizations and health departments can reach unscreened people (Brenner et al, Cancer, 2018)





### **CDC National Cancer Conference (August 2017)**

USING INDIVIDUAL-BASED SIMULATION MODELING TO INTEGRATE BIG DATA AND INTERVENTION EVIDENCE TO INFORM INTERVENTION SELECTION, ADAPTATION, AND EVALUATION: AN EXAMPLE ON COLORECTAL CANCER SCREENING

PRESENTED BY: KRISTEN HASSMILLER LICH, PHD MHSA ASSISTANT PROFESSOR, UNIV OF NORTH CAROLINA AT CHAPEL HILL



#### Relationships, Data, and Quality Improvement Infrastructure

Critical Factors when Accountable Care Organizations and Primary Care Practices Collaborate to Increase Colorectal Cancer Screening in Medicaid Members

PRESENTED BY: Melinda M. Davis, PhD, Director of Community Engaged Research, Associate Professor – Department of Family Medicine DAT: August 14, 2017 LOCATION: CDC National Cancer Conference, Atlanta, MD





#### Data-powered decision making: One state's approach to improving colorectal cancer screening in underserved populations

#### Stephanie B Wheeler, PhD MPH

- Associate Professor
- University of North Carolina at Chapel Hill



Technical Considerations: the past, present and future of simulation modeling of colorectal cancer

INDUSTRIAL AND SYSTEMS ENGINEERING

Siddhartha Nambiar, Rachel Townsley, Maria Mayorga North Carolina State University

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### **North Carolina CRC Simulation Model**

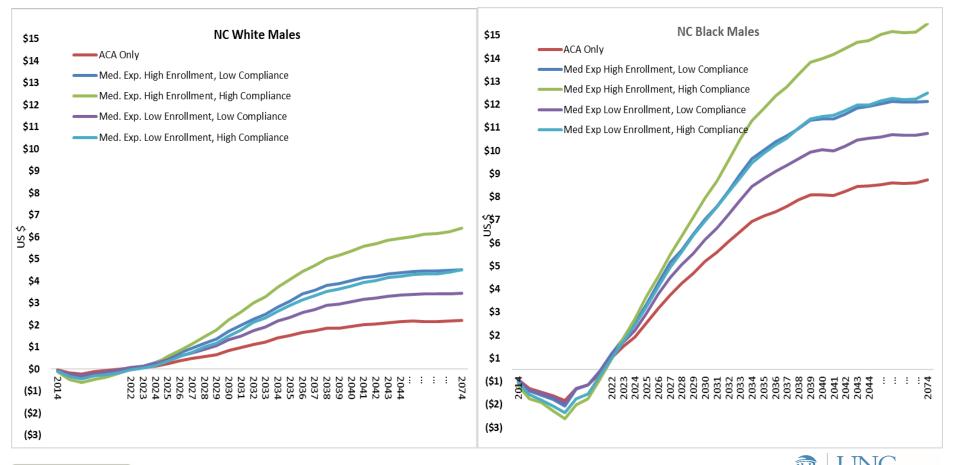
Simulated age-eligible NC population up to date with CRC screening on January 1, 2023

Variable	No ACA	Percentage point change in percent up-to-date on CRC screening compared with the removal of ACA		
		ACA	ACA + Medicaid Expansion	
Overall	48.65%	+1.03%	+1.74%	
By sex				
Male	46.13%	+0.94%	+1.55%	
Female	51.00%	+1.11%	+1.92%	
By race				
White	49.92%	+0.73%	+1.29%	
Black	45.92%	+2.01%	+2.88%	
Hispanic	42.22%	+0.05%	+2.90%	
Other	42.36%	+1.40%	+3.40%	
By insurance				
Private	53.87%	+0.01%	+0.03%	
Dual	58.02%	+0.02%	+0.99%	
Medicare	59.85%	+0.09%	+0.15%	
Medicaid	42.63%	+0.07%	+0.02%	
Uninsured	17.84%	-0.04%	-0.04%	





## Medicaid expansion is expected to be cost-saving in terms of CRC screening and outcomes over time





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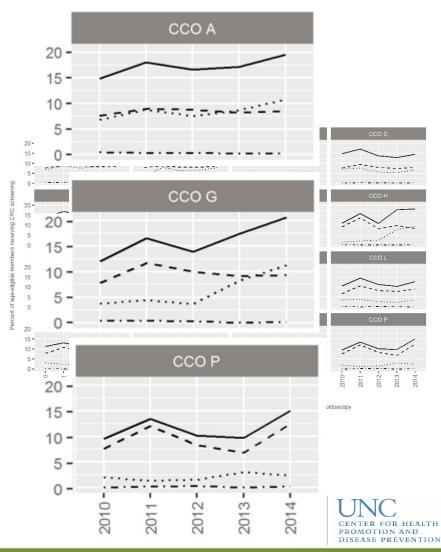
PROMOTION AND

### CRC Testing among Oregon Medicaid Enrollees, 2010-2014

- **Objective:** To assess CRC testing patterns statewide and by Medicaid coordinated care organizations (CCOs)
- **Population:** 134,424 Oregon Medicaid members ages 50-64

#### Results:

- Probability of CRC testing increased by 0.7 percentage points in 2011 and 1.4 percentage points in 2014 (versus 2010)
- 3-fold increase in fecal testing in 2014 compared to earlier years
- Modality patterns vary by CCO





### CRC Screening Among 50-Year-Old Oregon Medicaid Enrollees

- **Objective:** To determine if Medicaid members are differentially likely to get screened based on the year in which they turn 50 and/or enroll in Medicaid, and to map these patterns onto state and federal policies
- **Population:** 14,576 Oregon Medicaid enrollees who turned 50 from 2010 to 2014
- Results:
  - Individuals who enrolled in Medicaid for the first time in 2013 (RR: 1.58; 95% CI=1.20, 2.09) or 2014 (RR: 1.31; 95% CI=1.15, 1.49) were more likely to get screened than those enrolled in 2010
  - Having a primary care visit in the calendar year, chronic disease, and being Hispanic were also significantly more likely to be screened





### **Oregon Medicaid and Clinic Partnerships**

- **Method:** qualitative comparative study of 14 Oregon CCOs and their regional primary care clinics
- **Data Sources:** public performance data, key informant interviews, consultation field notes

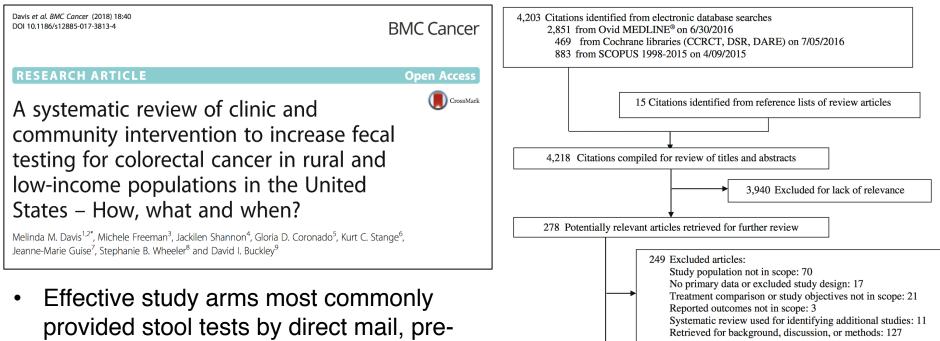
#### • Themes:

- 3 key partnership dimensions:
  - Establishing relationships and building partnerships
  - Producing and sharing performance data
  - Developing a process and infrastructure to support quality improvement
- 2 unintended consequences:
  - Potential exclusion of smaller clinics
  - Metric fatigue





### Fecal Testing Interventions in Rural and Low-Income Populations



- addressed stamped envelopes, client reminders, and in-clinic distributions.
- More guidance is needed regarding which interventions work best for specific settings, populations, and community characteristics.





27 Primary studies published in 29 articles

23 Controlled trials: 21 Randomized

2 Non-randomized

1 Feasibility assessment

4 Other study designs: 3 Pre-post

### **Opportunities to Improve Rural Cancer Control**

THE JOURNAL OF R	URAL	HEALTH	NRHA			
COMMENTARY <b>"Taking the Bull by the Horns": Four Principles to Align Public</b>						
Health, Primary Care, and Community Efforts to Improve Rural Cancer Control						
Stephanie B. Wheeler, PhD, MPH <sup>1</sup> & Melinda M. Davis, PhD <sup>2</sup>						
The Journal of Rural Health <b>33</b> (2017) 345–349 (	© 2017 Nationa	al Rural Health Association				

- 1. Utilize existing data when possible and develop new methods for working with small sample sizes.
- 2. Prioritize efforts to evaluate, adapt, and expand EBIs to rural areas using multidisciplinary research strengths.
- 3. Weigh the pros and cons of rural definitions and consider the interaction of geography with individual-level and regional factors.
- 4. Utilize an equity-based participatory implementation science approach to improve and align research and quality improvement efforts.



