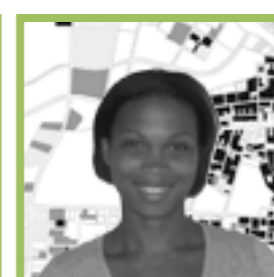
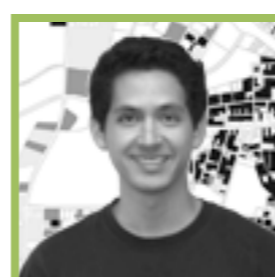
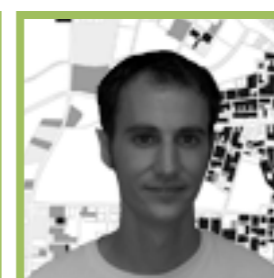
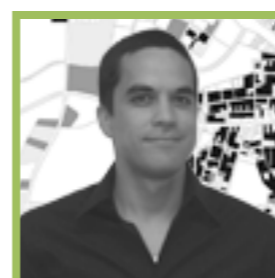
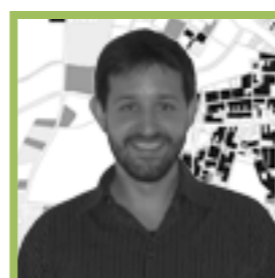
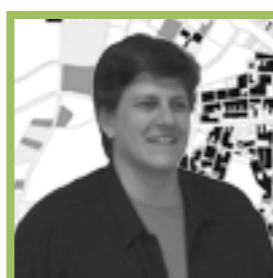


# Flexible Models for Elucidating Health Disparities

by Eric C. Tassone, J.D., Ph.D., Alan E. Gelfand, Ph.D., & Marie Lynn Miranda, Ph.D.



- Focus on children
  - Focus on issues of environmental justice
    - Shift to preventive interventions
      - Emphasis on spatial analytic approaches





# Outline of Talk

- **Overview: “What”, “how”, and “why” of disaggregation**
- **Modeling Details**
- **Computing Details and Issues**
- **Example: North Carolina Detailed Birth Record Data**
- **Impact on Policy?**
- **Future Directions**



- **Disaggregation? What, how, and why...**
  - **What:**
    - **A method that substantially extends inferential possibilities of customary modeling of spatial outcomes data that are areally observed**
    - **Computationally tractable for large data sets (on the order of  $10^5$  or more)**
    - **'Disaggregated': we disaggregate aggregated counts from usual spatial model into subgroups using individual-level characteristics**



- **Disaggregation? What, how, and why...**
  - **How:**
    - **Model subgroups in areal unit using individual-level data**
    - **Multi-way contingency table for each areal unit**
    - **Explained with loglinear model in each areal unit**
    - **Spatially smooth models via random effects**
    - **All in a multilevel modeling framework**



- **Disaggregation? What, how, and why...**
  - **Why?:**
    - **Other approaches problematic with subgroups**
      - **Sometimes adjust via covariates or expected**
      - **How to model the subgroups? ANOVA-like approach? Ind. models? Via multivariate CAR priors? Ignore / aggregate?**
      - **Confined to 'outcome' cond. on 'risk factors'**
    - **Use available individual-level data..true level?**



- **Disaggregation? What, how, and why... more why...**
  - **Flexible inference in multilevel structure**
  - **Dimension reduction ( $r \ll L$  in general)**
  - **No need to specify a “response” variable**
  - **Joint modeling=>arbitrary marginal and cond. probs.**
    - **Not just conditional probability statements**
    - **Arbitrary marginal, joint, and cond. statements**
    - **Flexible aggregation: investigate outcomes/ groups of interest, e.g., racial disparities**

Cell counts:

$$n_l^{(s)} \sim Po(\lambda_l^{(s)})$$

First level:  $\log(\lambda_l^{(s)}) = \mathbf{X}_l^T \beta_s + \log(n_l^{(s)}) = \sum_{t=1}^r X_{lt} \beta_{st} + \log(n_l^{(s)})$

Second level:  $\beta_{st} = \mathbf{w}_s^T \eta_t + \tilde{\phi}_t^{(s)} = \sum_{u=1}^q w_{su} \eta_{tu} + \tilde{\phi}_t^{(s)}$

Random effects:  $\tilde{\phi}_t^{(s)} = \phi_t^0 + \phi_t^{(s)}$

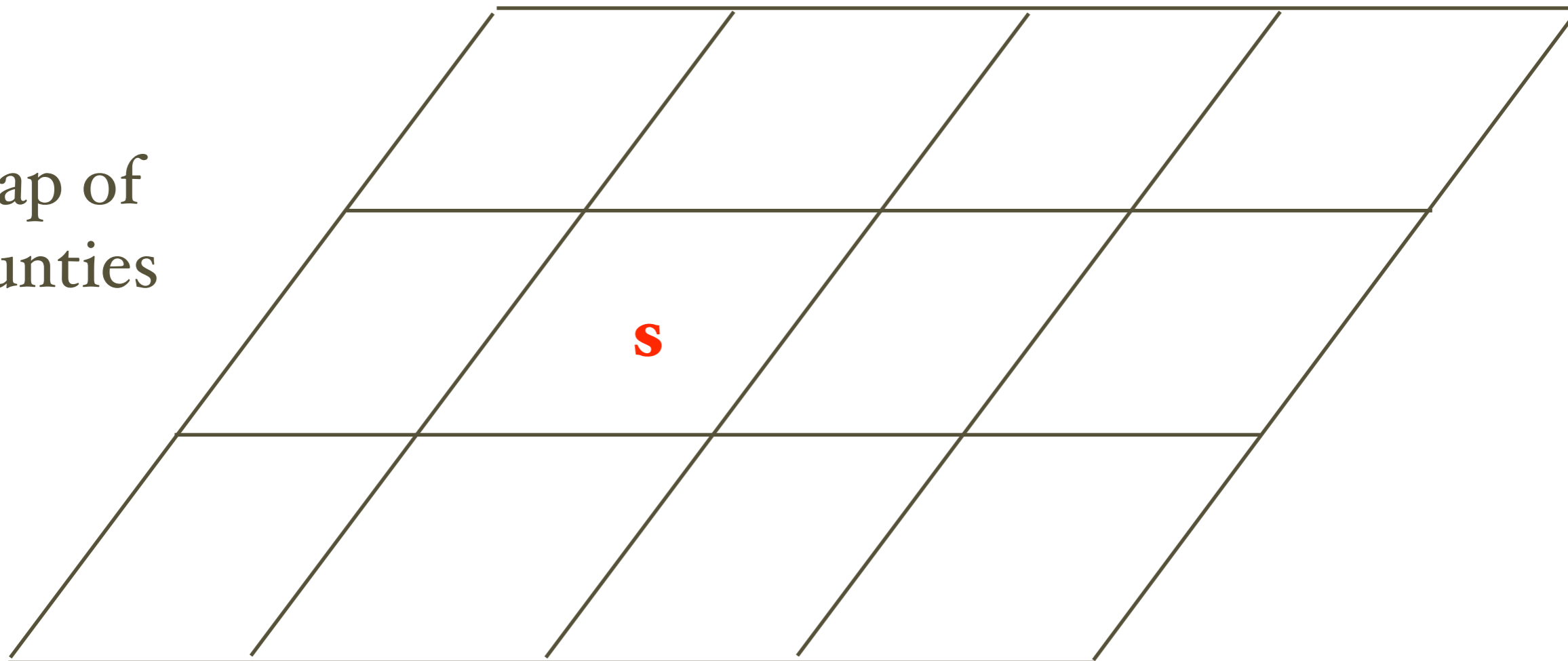
And we can plug the second level into the first level for very nice interpretations of the overall model, namely the log counts as...

$$\mathbf{X}\phi^0 + \mathbf{X}\mathbf{W}_s\eta + \mathbf{X}\phi^{(s)} .$$

(+ offset)

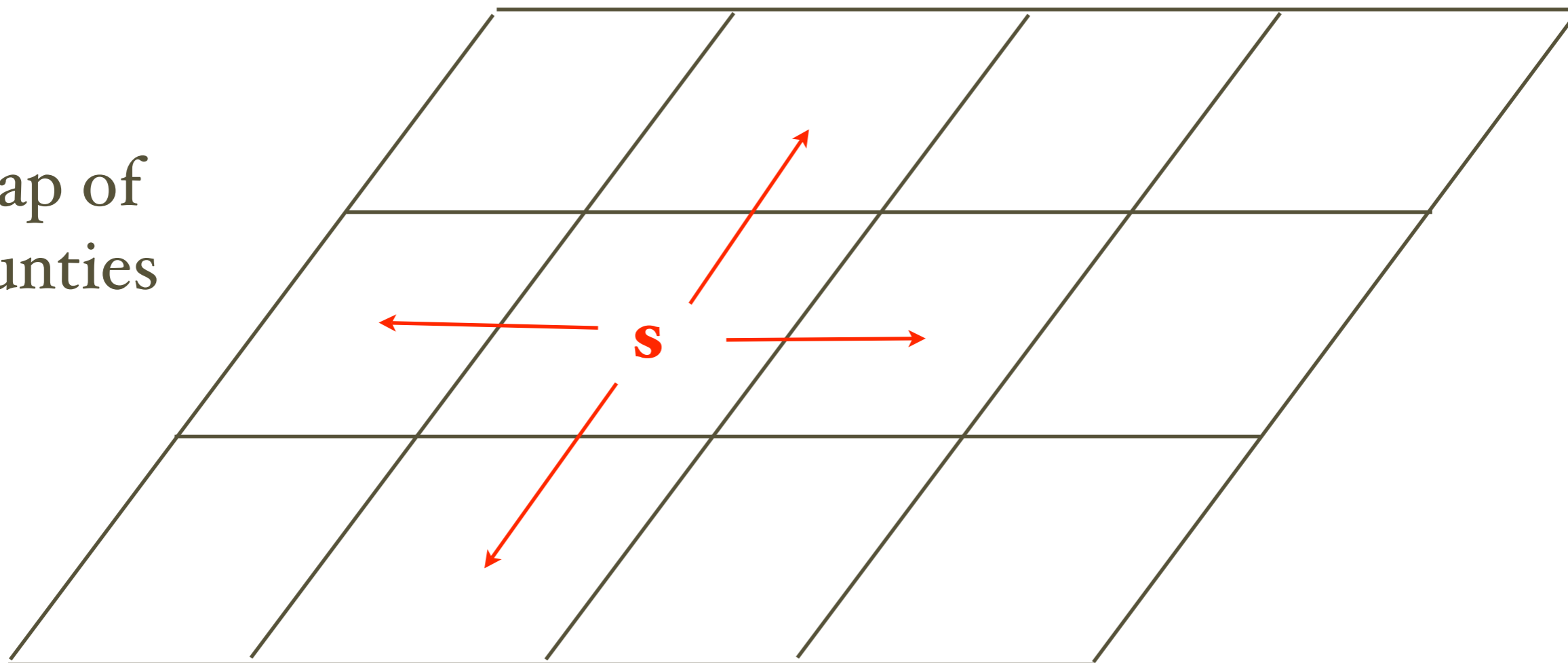


Map of  
counties



Smooths county **s**'s  $\beta_{st}$  to be like its neighbors (for selected  $t$ 's).

Map of  
counties



Smooths county **s**'s  $\beta_{st}$  to be like its neighbors (for selected  $t$ 's).



# Computing Details

- Can be implemented in WinBUGS ... computationally OK
- Many modeling choices in this flexible framework...
  - Design matrix for loglinear model...
  - Which loglin. model parameters get spatial smooth?
  - Which get areal unit-level covariates? Which covars.?
  - What form for spatial random effects?
    - Independent?
    - Multivariate CAR?
    - Attempt some sort of dimension reduction?
- For the example we'll see, we will detail our choices



- **NC Detailed Birth Record**
  - **1999-2003**
  - **No congenital anomalies**
  - **Singletons**
  - **N=463,639, with 32,437 LBW (~6.996%)**
- **County-level (though finer resolution available)**
  - **ZCTA ... alternative state analysis?**
  - **Census Tract, Block Group, Block...focused analysis?**



## Example

- **Variables**

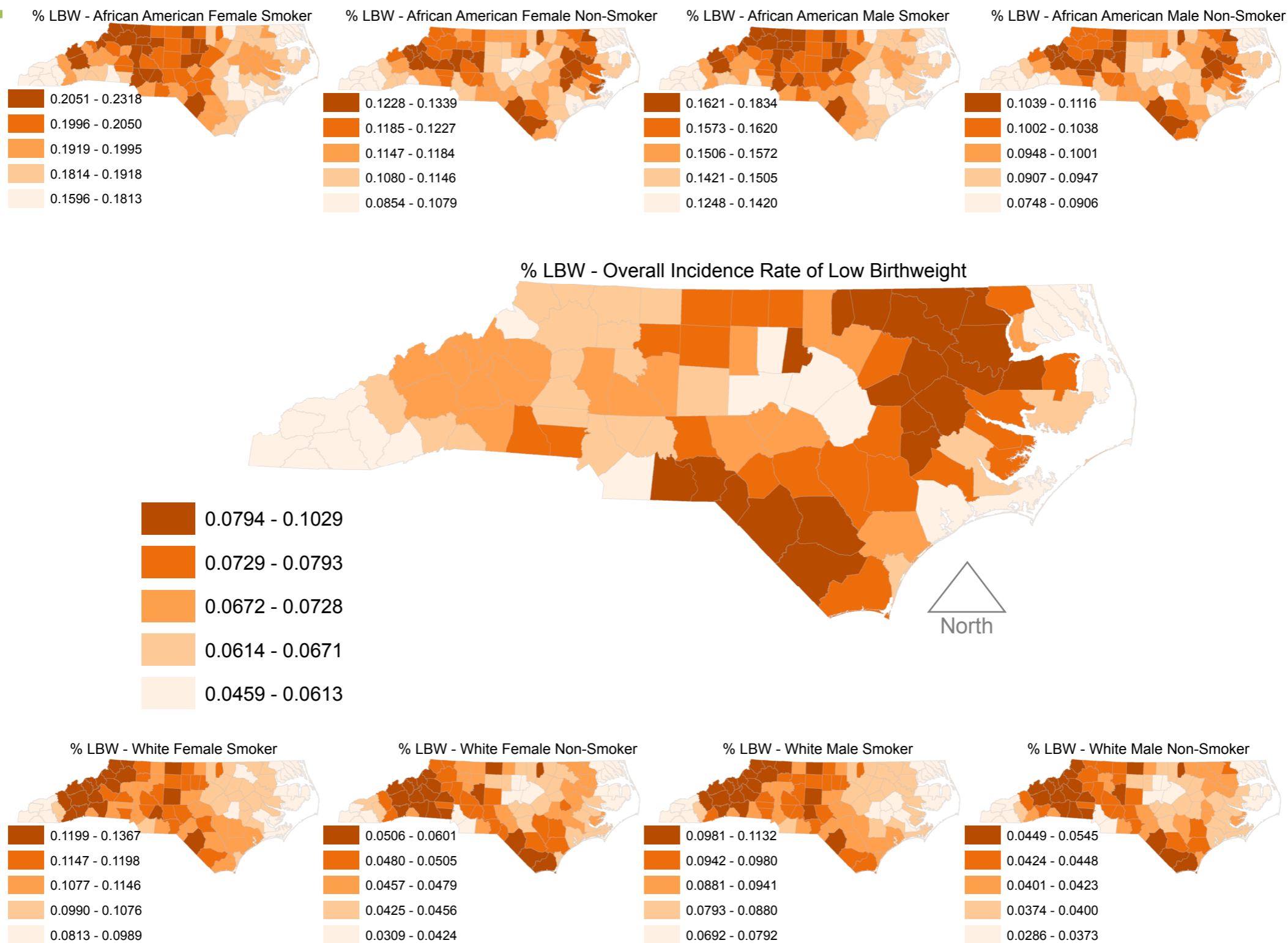
- **X: Maternal race (African Amer. (AA) or white)**
- **Y: Low birth weight (yes or no)**
- **Z: Sex of infant (female or male)**
- **W: Maternal tobacco use (yes or no) - “Smoking”**
- **So  $2 \times 2 \times 2 \times 2 = 16$  subgroups in each areal unit**

- Let  $L$  denote subgroup,  $l=1, 2, \dots, 16$
- Let  $S=1, 2, \dots, 100$  index counties in NC
- Model provides  $p_l^{(s)}$ , the probability for subgroup  $l$  in county  $s$  ... not prob. LBW given  $X, Z, W$ , but joint prob..
  - Re-combine into cond., marginal, etc., probs.
  - Disparity measures (e.g., odds ratio)
- Model fit:  $(XYZ, XYW, YZW)$ 
  - So reduce from 16 to 14 ... more reduction in general

# Example: Model

$$\begin{aligned}
 \log \lambda_{ijklm}^{(s)} = & \gamma^{(s)} \\
 & + \gamma_i^{X(s)} + \gamma_j^{Y(s)} + \gamma_k^Z + \gamma_m^{W(s)} \\
 & + \gamma_{ij}^{XY(s)} + \gamma_{ik}^{XZ(s)} + \gamma_{im}^{XW(s)} + \gamma_{jk}^{YZ(s)} + \gamma_{jm}^{YW(s)} + \gamma_{km}^{ZW(s)} \\
 & + \gamma_{ijk}^{XYZ} + \gamma_{ijm}^{XYW} + \gamma_{jkm}^{YZW} + \log(n_{\cdot}^{(s)}).
 \end{aligned}$$

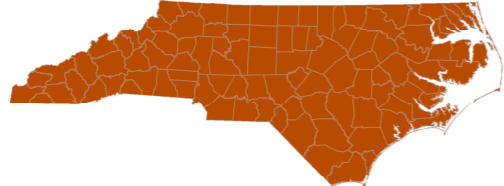
# Est. LBW%, Overall (center) and Subgroups



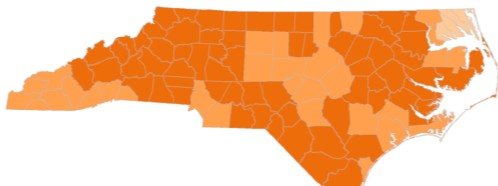


# Est. LBW%, Overall (center) and Subgroups

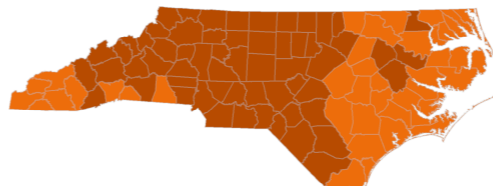
% LBW - African American Female Smoker



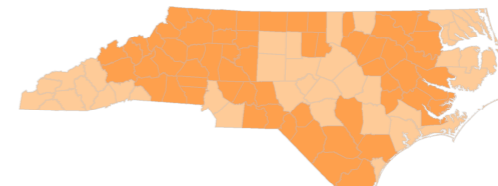
% LBW - African American Female Non-Smoker



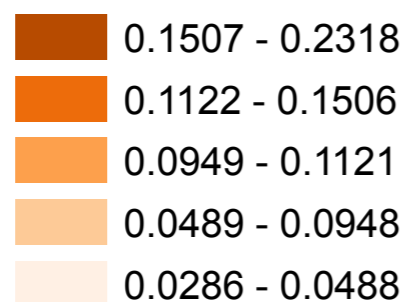
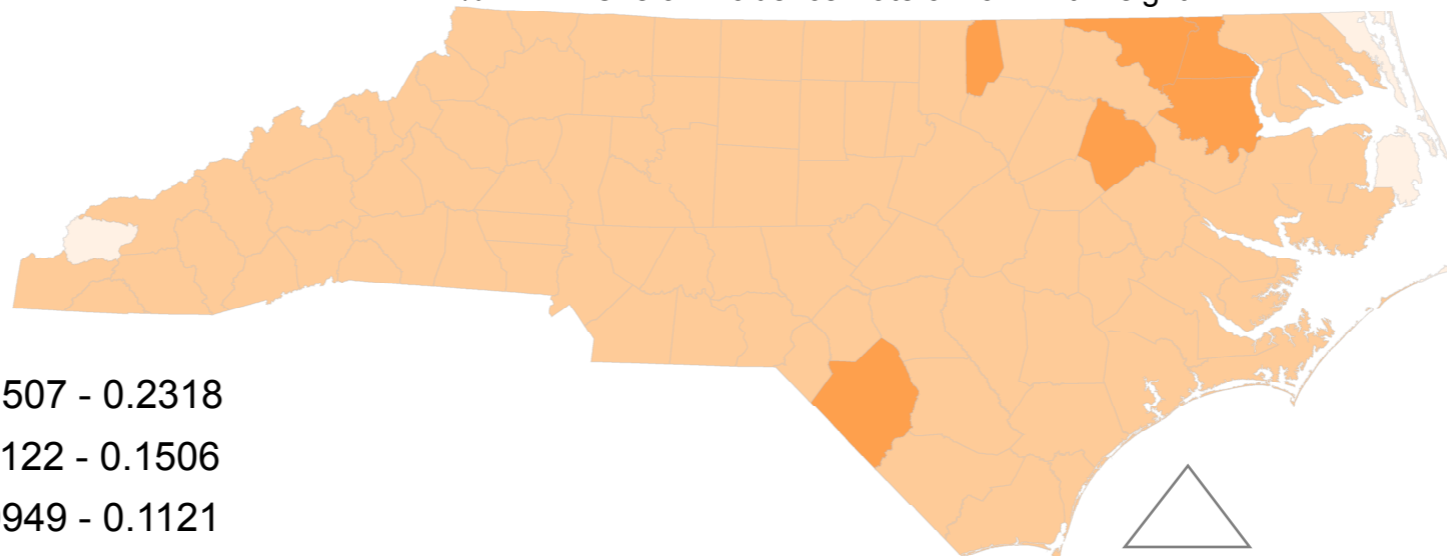
% LBW - African American Male Smoker



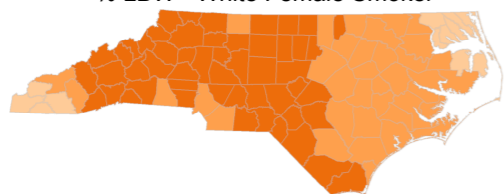
% LBW - African American Male Non-Smoker



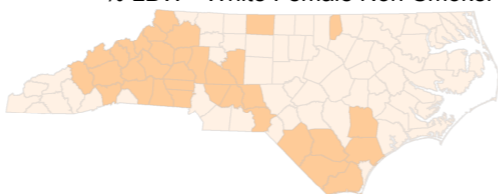
% LBW - Overall Incidence Rate of Low Birthweight



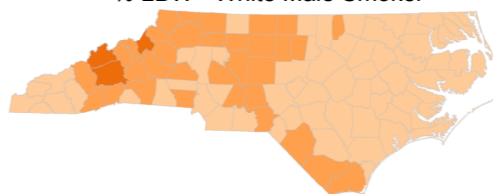
% LBW - White Female Smoker



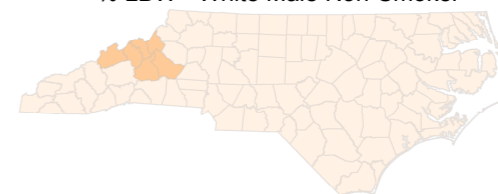
% LBW - White Female Non-Smoker



% LBW - White Male Smoker

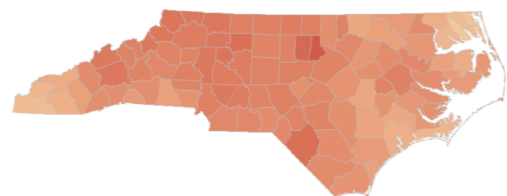


% LBW - White Male Non-Smoker

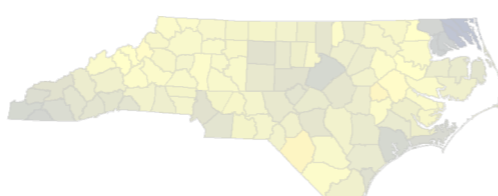


# Est. LBW%, Overall (center) and Subgroups

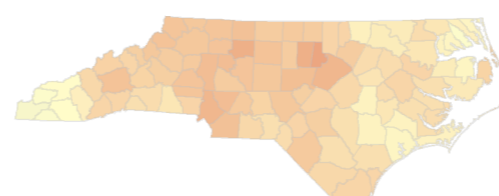
Girl, AA Smoker



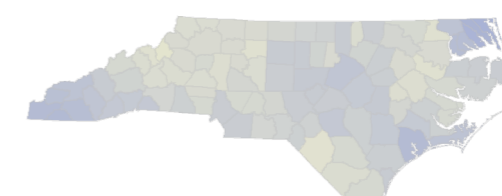
Girl, AA nonSm



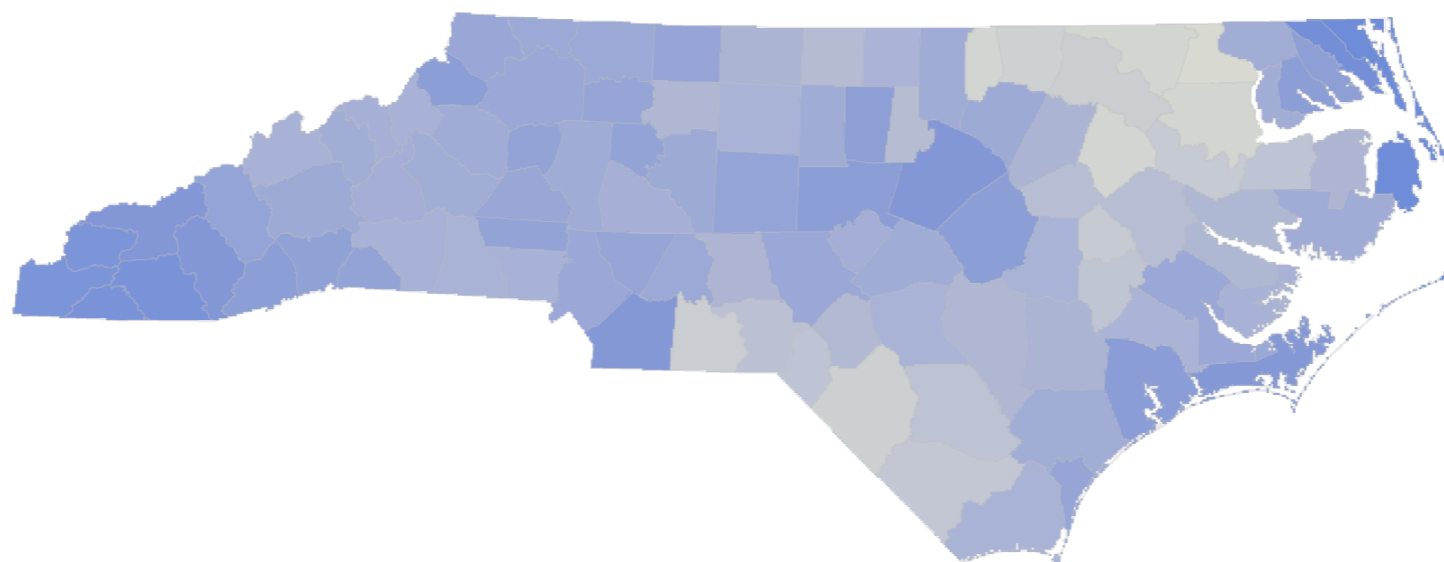
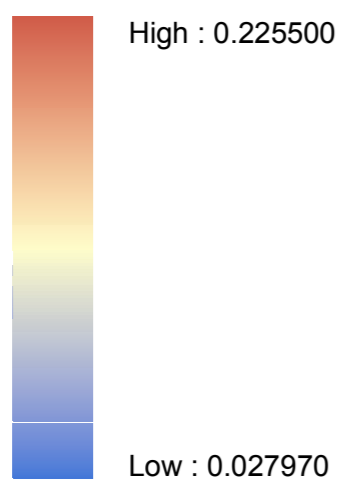
Boy, AA Smoker



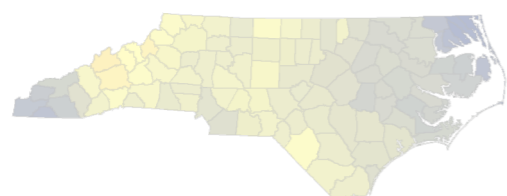
Boy, AA nonSm



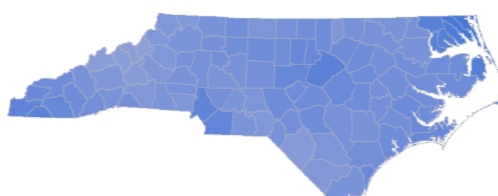
All births



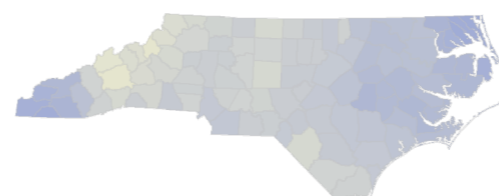
Girl, W Smoker



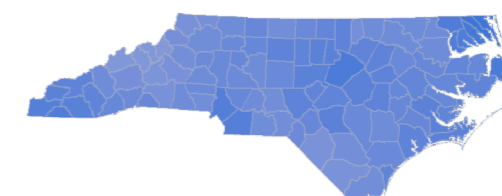
Girl, W nonSm



Boy, W Smoker

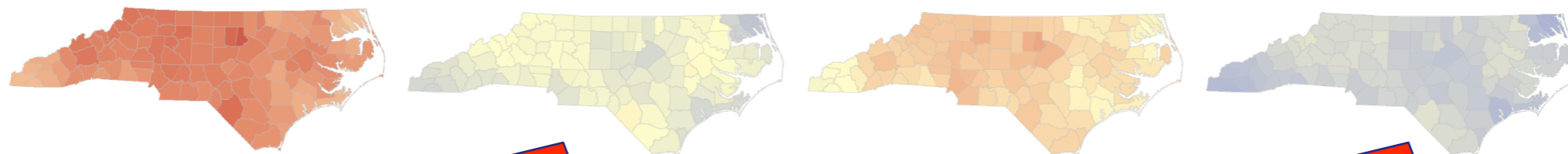


Boy, W nonSm

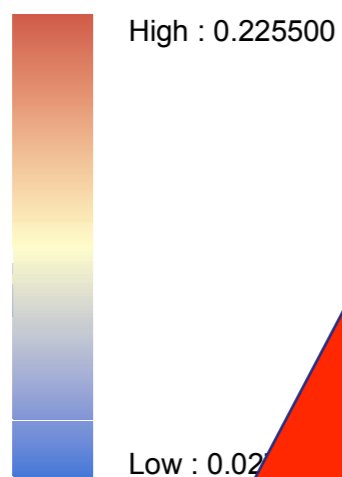


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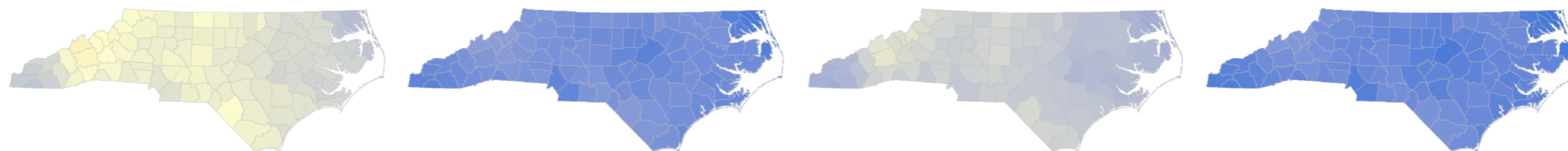
Girl, AA Smoker      Girl, AA nonSm      Boy, AA Smoker      Boy, AA nonSm



All births

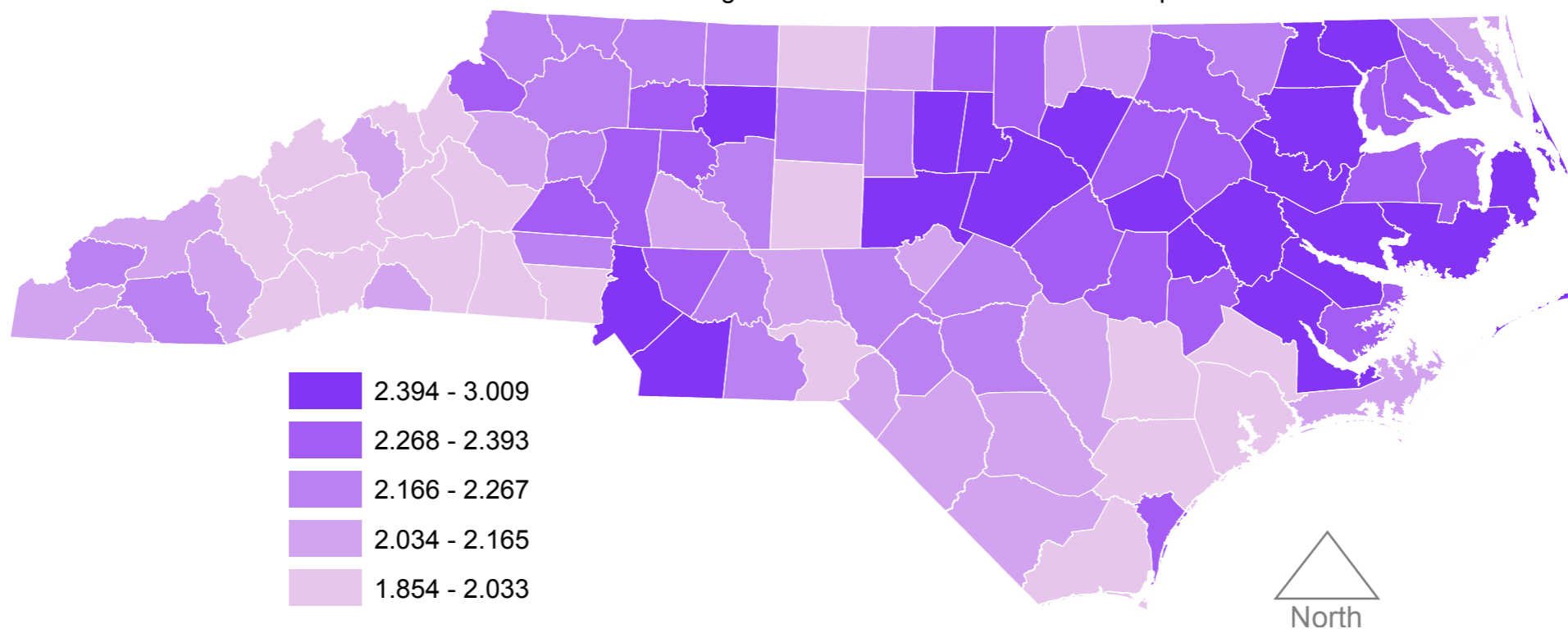


Girl, W Smoker      Girl, W nonSm      Boy, W Smoker      Boy, W nonSm



# Odds Ratio for Race

Odds ratio of low birth weight for African American mothers compared to white mothers



We can think of this as a (relative) measure of racial disparity.



# Impact on Policy?

- **Healthy People 2010 (HP2010)**
  - **“Eliminate Health Disparities”**
    - **One of two overarching goals**
- **HP2010: Health disparities are “differences that occur by gender, race or ethnicity, education or income, disability, geographic location, or sexual orientation.”**
- **Today’s example:**
  - **(1) gender (i.e., subgroups),**
  - **(2) race (i.e., subgroups), and**
  - **(3) geographic location (i.e., spatial)**
    - **... and combines, in local disparity measures**



## Impact on Policy?

- Also, contextual effects via multilevel model framework
  - E.g., does 'individual' effect of race differ in areas w/ different socioeconomic or demographic features
- Measurement of health disparities
  - “Methodological Issues in Measuring Health Disparities”, NCHS (2005)
  - Work of Harper and Lynch
- “Methodological Issues...” emphasize both absolute and relative measures of disparity.
  - OR for race, shown above, is relative measure...
  - ...but flexible methods such as ours can easily accommodate both in some model
  - Also estimates component rates simultaneously

# Impact on Policy?

Helpful to see disparity in the context of the component rates...

## LBW, all and subgroups

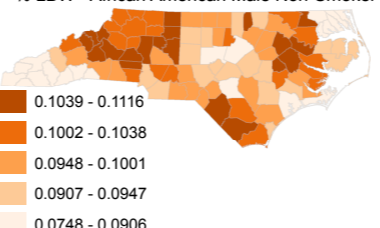
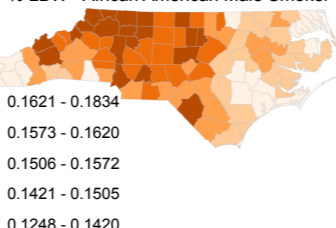
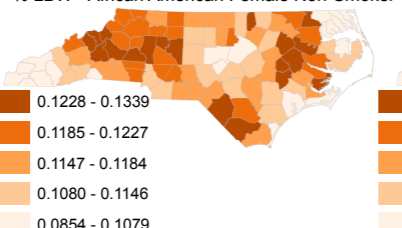
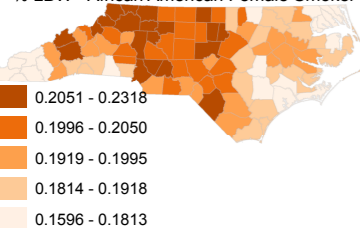
## OR race

% LBW - African American Female Smoker

% LBW - African American Female Non-Smoker

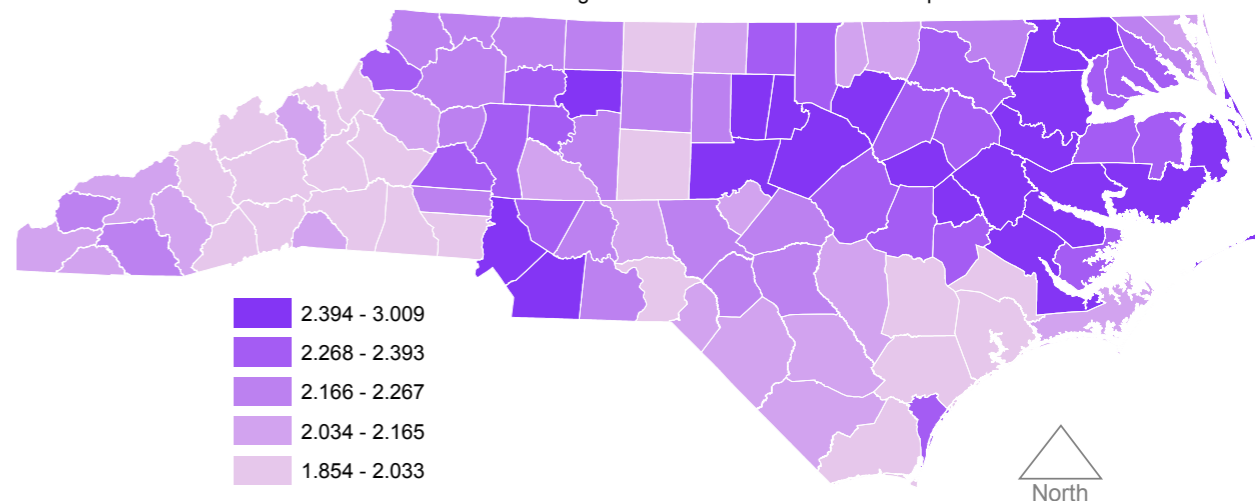
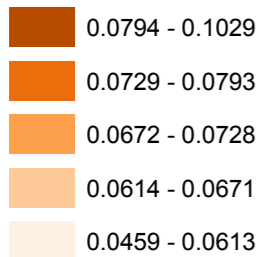
% LBW - African American Male Smoker

% LBW - African American Male Non-Smoker



% LBW - Overall Incidence Rate of Low Birthweight

Odds ratio of low birth weight for African American mothers compared to white mothers

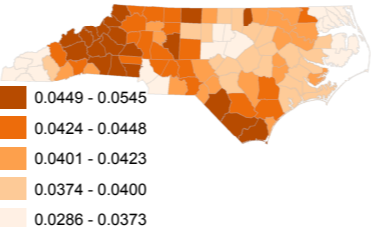
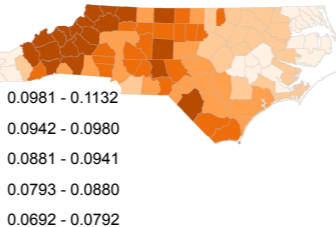
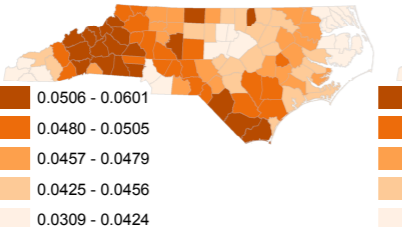
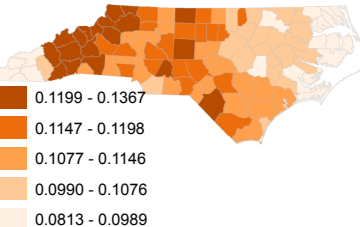


% LBW - White Female Smoker

% LBW - White Female Non-Smoker

% LBW - White Male Smoker

% LBW - White Male Non-Smoker





## Impact on Policy?

- **How might this inform policies?**
  - **Different priorities might follow from different relative or absolute disparity... which is higher priority?**
  - **Different “types” of disparity might suggest different interventions...**
    - **How does “two subgroups doing relatively well, but disparity high” compare to “both subgroups doing relatively poorly but disparity low”?**
    - **Focused intervention in area where one group doing relatively well but other relatively poorly?**
  - **Alternative measures of disparity? Excess deaths?**
  - **Proper measuring of interventions...what if both groups rate's go down, but disparity worsens?**  
**Suggests ‘disparity only’ measures miss something...**





# Future Directions

- **More individual-level data / more categories**
  - E.g., maternal education, an ordered categorical variable
  - Logic functions to reduce dimension
- **More contextual variables (areal unit covariates)**
  - income, demographics, etc., ...
  - interact w/ ind. level variables? Variables “with themselves” ...



# Future Directions

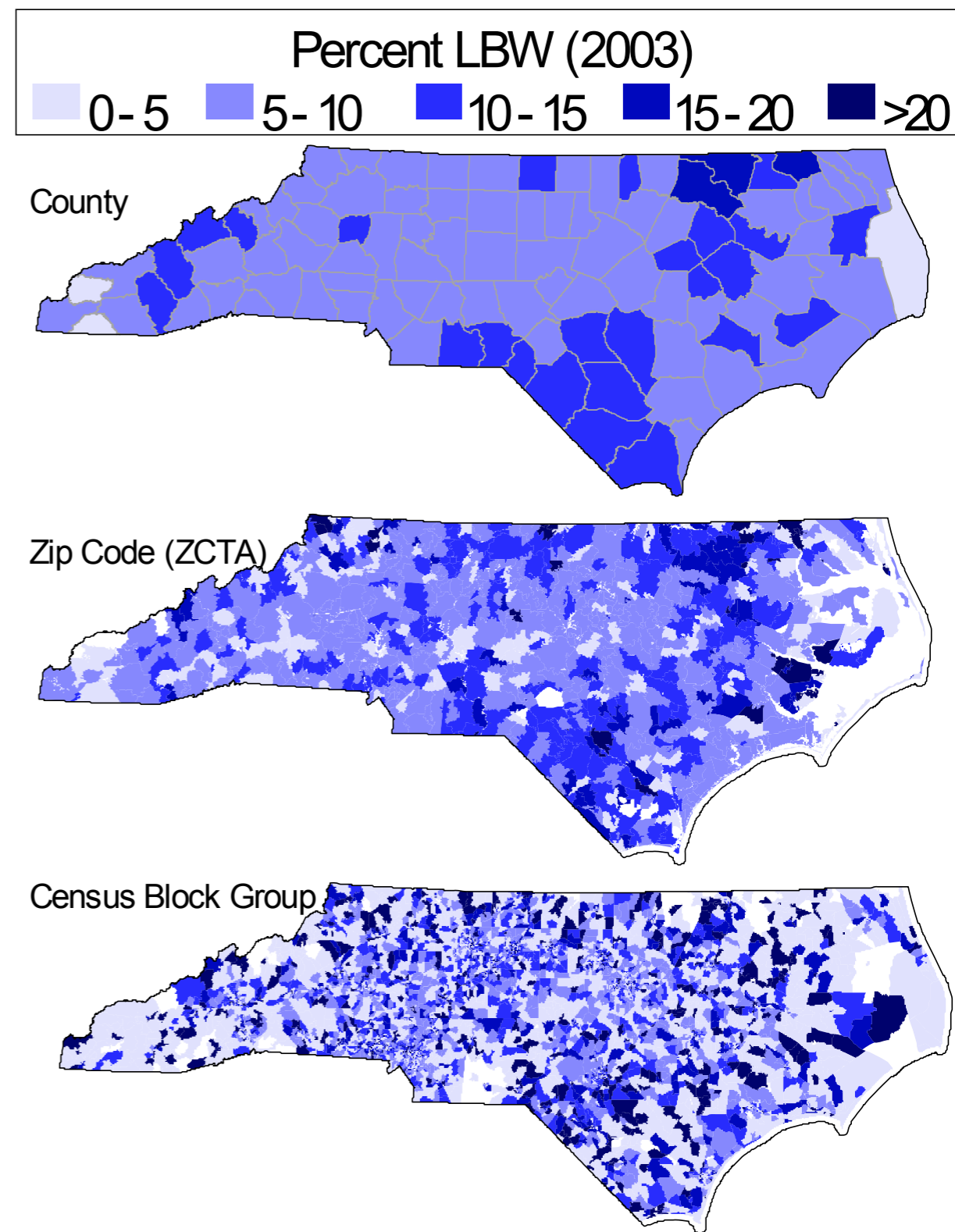
- Spatial scale
  - Now: Counties,
  - Next: ZCTA and beyond...
- Spatial loglinear modeling for point data...

- Spatial scale

- Now: Counties,
- Next: ZCTA and beyond...

**Figure 1.** Spatial pattern in percent of low birthweight births in North Carolina.

- Spatial loglinear modeling for point data...



- **Thank you very much!**