CS&SS 569 Visualizing Data and Models Lab 5: Intro to tile

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Overview of tile

Preview of three examples

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Scatterplot: HW1 example

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 - Expected probabilities and first differences: Voting example

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Installing tile and simcf

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- Preview of three examples
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 - Expected probabilities and first differences: Voting example
 - Ropeladder: Crime example
- Installing tile and simcf
- Walking through examples

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 - More later

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 - Could be a set of points, or text labels, or lines, or a polygon
 - Could be a set of points and symbols, colors, labels, fit line, Cls, and/or extrapolation limits
 - Could be the data for a dotchart, with labels for each line
 - Could be the marginal data for a rug
 - All annotation must happen in this step
 - Basic traces: linesTile(), pointsile(), polygonTile(), polylinesTile(), and textTile()
 - Complex traces: lineplot(), scatter(), ropeladder(), and rugTile()

Primitive trace functions:

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 - scatter(): Plot scatterplots with text and symbol markers, fit lines, and confidence intervals

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 - Set up the rows and columns of plots
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 - 3. **Examine output and revise**: Look at the graph made in step 2, and tweak the input parameters for steps 1 and 2 to make a better graph

Three examples

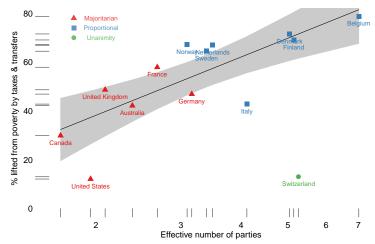


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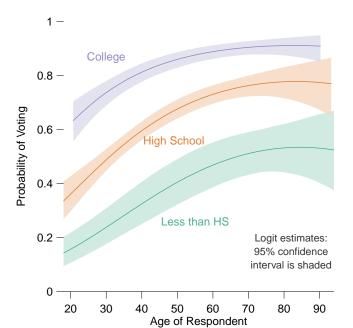
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Party Systems and Redistribution



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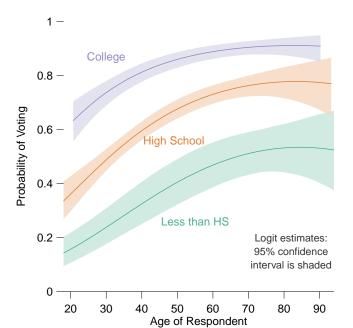
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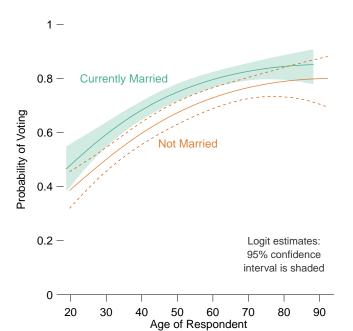
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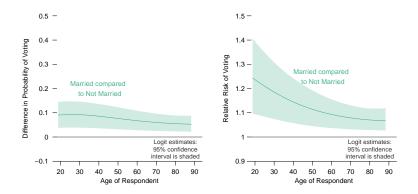
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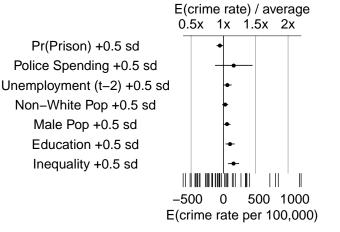
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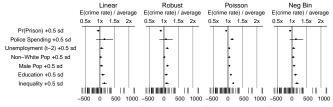
- 4. Simulate quantities of interest by compounding those 10,000 $\tilde{\beta}_k$ with counterfactual scenarios
 - Then compute average (point estimate) and appropriate percentiles (confidence intervals)
 - simcf::logitsimev() for expected values for logit models
 - logitsimfd for first differences
 - logitsimrr for relative risks



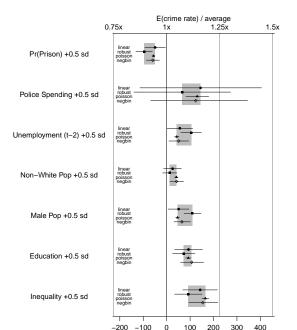


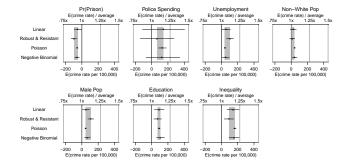






E(crime rate per 100,000)E(crime rate per 100,000)E(crime rate per 100,000)E(crime rate per 100,000)





Installing tile and simcf

- Go to Chris's website, Software section
- Also download all R scripts and data under today's Lab section