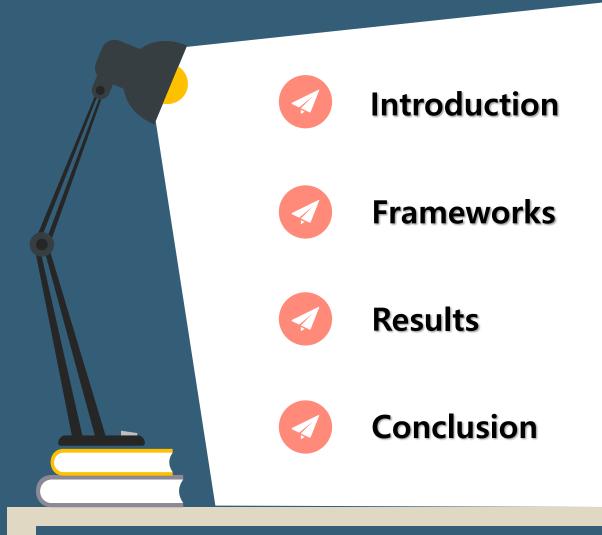


Research Paper

Regional disparity in the uninsurance rate impact of COVID-19: a spatial machine learning approach

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Introduction

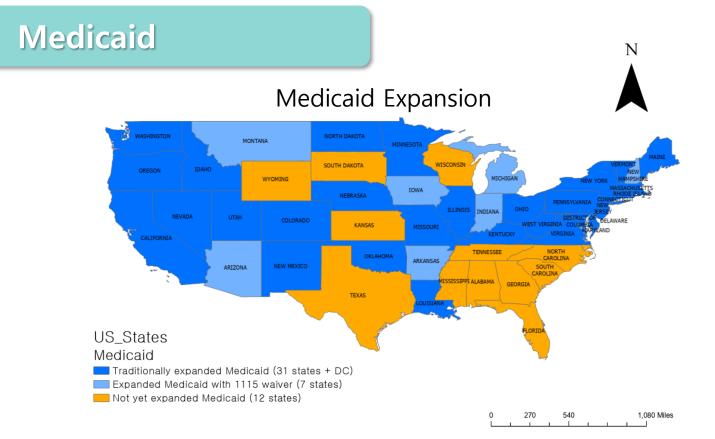
Introduction



- The primary form of private insurance in the United States is Employer-Sponsored Insurance (ESI).
- Medicare and Medicaid subsidize the health insurance system to a limited extent.
- Most Americans cover their medical problems with private health insurance.



- ► The problem is that COVID-19 has caused massive unemployment in American society, which has increased the uninsured rate.
- ► The US government and each state government have eased Medicaid enrollment standards and simplified procedures to reduce the damage caused by COVID-19.



Source: National Library of Medicine

- Medicaid is a public health insurance program in the United States that provides low-income people ‡ under the age of 65, children under the age of 19, pregnant women, and people with disabilities.
- It is jointly financed by the federal and state governments and operated by the state governments.
- To mitigate the impact of COVID-19, several states have expanded Medicaid standards.
- However, discussions on the effect of Medicaid expansion are still conflicting.

Medicaid Expansion

Positives for Medicaid Expansion

Soni, A. (2020); Creedon et al. (2022); Bellerose, Collin, & Daw. (2022); Dong, Gindling, & Miller (2022).

- Reduce ethnic disparities in fertility and preterm birth rates.
- Increase investment in healthy behaviors among low-income people.
- Increase flu vaccination rates.
- Medicaid expansion reduces uninsured, privately insured rates.

Skepticism over Medicaid Expansion

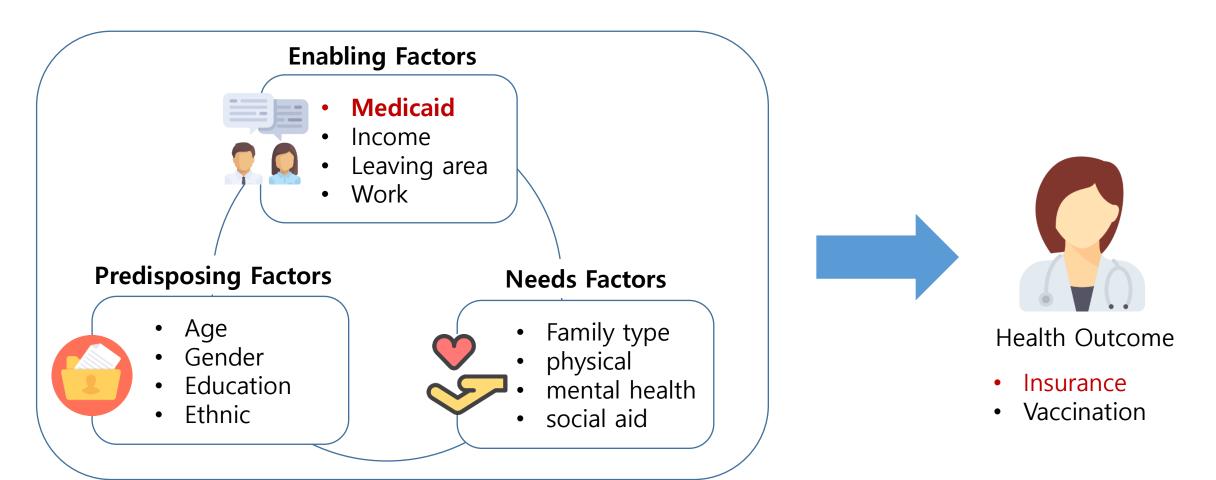
Ghosh, Simon, & Sommers (2019); Takvorian et al. (2020); Kandilov, & Kandilov (2022)

- No evidence of an improvement in the appropriate treatment rate was found.
- The effect varies by state, county, and region.
- It does not affect the private insurance subscription rate according to the type of work.
- ▶ The empirical evidence of decreasing uninsured rates is inconsistent across regions.

Anderson Behavior Model

Andersen & Newman, 1973; Van Houtven & Norton, 2004

• This model is an analytical framework that predicts medical behavior by considering internal and external factors.



Research Goal

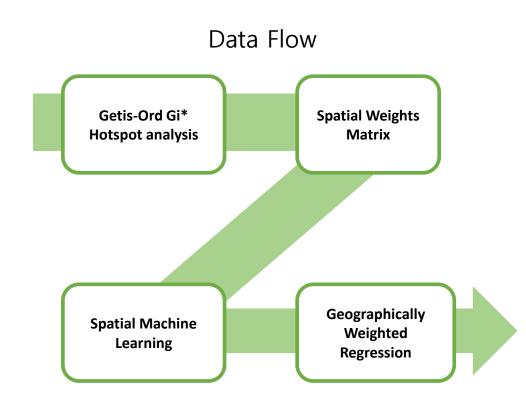
- Effects of Medicaid Expansion and Other Social Factors on Health Insurance Uninsured Rates.
- Existing statistical estimation methods are limited in considering various social factors simultaneously and cannot predict changes in the importance of each variable in the future.
- Therefore, this study creates the importance index between variables, and a prediction model using machine learning approaches.
- Whether there is an influence of the spatial patterns of the counties.
- In various preceding studies, social variables, including Medicaid, have different influences depending on regional characteristics. Therefore, I want to reflect on the spatial features in the model.

Research Design

Research Design

- Unit of Analysis: US counties (3,108).
- Research time scope: 2019, 2020, and 2021.
- County Health Ranking Data (the University of Wisconsin's Population Health Institute)

- Gi* Hotspot show the county's spatial autocorrelation and regional patterns.
- Spatial ML shows the importance of variables and predictive models based on spatial weights matrix.
- GWR illustrates the coefficient value for each county of the variable.

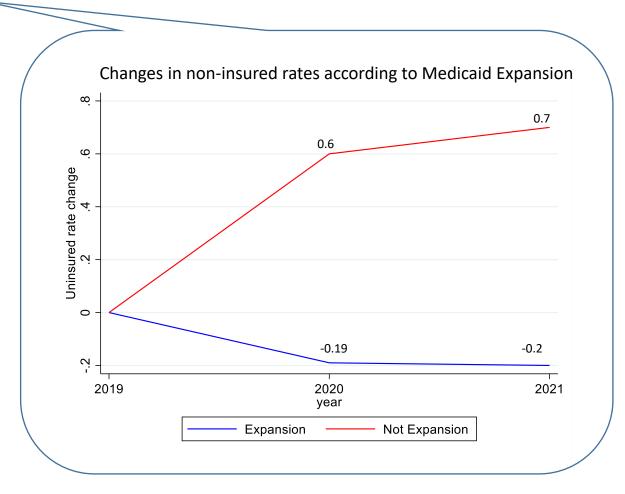


Data and Variables

| Variables | | | Descriptive | | |
|--------------------|-----------------------------|--------------------------------------|--|--|--|
| Target Variable | Change of Uninsurance Ratio | | $Uninsurance \ Ratio_t - Uninsurance \ Ratio_0$ | | |
| Feature | Me | dicaid | Medicaid expansion during COVID19 | | |
| Variables | Enabling Factors | Income inequality | Ratio of household income at the 80th percentile to income at the 20th percentile | | |
| | | Unemployment | Percentage of population ages 16 and older unemployed but seeking work | | |
| | | Life expectancy | Life expectancy | | |
| | Needs | Adult smoking | Percentage of adults who are current smokers | | |
| | Factors | Adult obesity | Percentage of the adult that reports a BMI greater than or equal to 30 kg/m2 | | |
| | | Food | Index of factors that contribute to a healthy | | |
| | | environment | food environment, from 0 (worst) to 10 (best). | | |
| | | Children in single-parent households | Percentage of children that live in a household headed by a single parent | | |
| | | Social associations | Number of membership associations per 10,000 population | | |
| | Predisposing | Age ratio | Percentage of 65 and over | | |
| | Factors | Ethnic | Percentage of Hispanic | | |
| | | Gender | Percentage of Female | | |
| | | Education | Percentage of adults ages 25-44 with some post-secondary education | | |
| | | Population | Number of Population | | |

Descriptive statistics

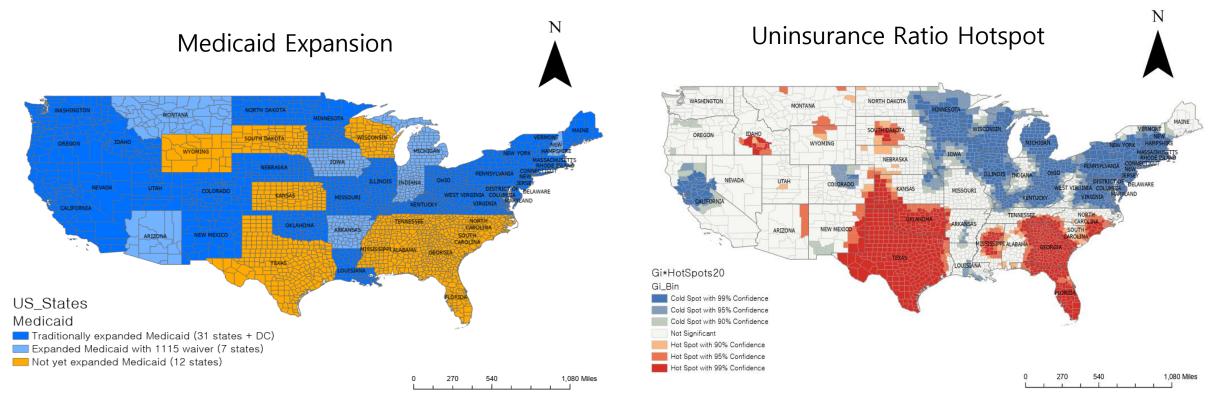
| Variables | | Before COVID | After COVID | |
|-----------------------|--------------------------------------|----------------|-----------------|-----------------|
| | | 2019 | Year 1 (2020) | Year 2 (2021) |
| Uninsurance Ratio (%) | | 11.05 | 11.41 | 11.45 |
| Medicaid (%) | | 63.4 (0.48) | 63.4 (0.48) | 63.4 (0.48) |
| Enabling Factors | Income inequality | 4.51 (0.74) | 4.51 (0.76) | 4.50 (0.77) |
| | Unemployment | 4.57 (1.56) | 4.09 (1.39) | 3.96 (1.39) |
| | Life expectancy | 75.84 (11.38) | 75.75 (11.71) | 75.89 (11.41) |
| Needs Factors | Adult smoking | 17.83 (3.57) | 17.44 (3.55) | 21.30 (4.11) |
| ractors | Adult obesity | 32.08 (4.58) | 32.85 (5.42) | 33.46 (5.94) |
| | Food environment | 7.43 (1.28) | 7.41 (1.26) | 7.41 (1.27) |
| | Children in single-parent households | 32.36 (10.53) | 32.32 (10.67) | 24.42 (10.00) |
| | Social associations | 13.81 (7.05) | 11.72 (5.88) | 11.62 (5.90) |
| Predisposing | Age ratio | 18.86 (4.56) | 19.31 (4.65) | 19.80 (4.75) |
| Factors | Ethnic | 9.51 (13.81) | 9.69 (13.90) | 9.81 (13.93) |
| | Gender | 49.92 (2.18) | 49.91 (2.23) | 49.91 (2.21) |
| | Education | 57.61 (11.67) | 57.89 (11.82) | 58.14 (11.93) |
| | Population | 104102(334347) | 104571 (334793) | 104920 (334768) |
| *The Observa | tions for all variab | les are 3,108. | | |



Getis-Ord Gi* Hotspot analysis

Global Moran's I
Moran's Index: 0.3729
Expected Index: -0.0003

Variance: 0.00003z-score: 66.411



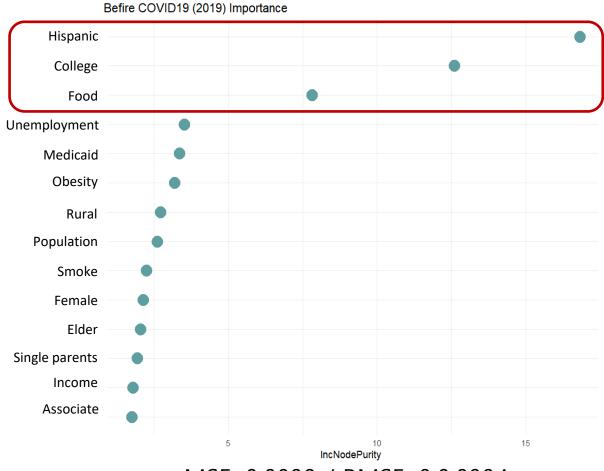
- Red: Areas with increased uninsured rates compared to surrounding areas
- Blue: Areas with reduced non-insurance rates compared to surrounding areas
- There are areas where Medicaid has been implemented, but the non-insurance rate has rather increased, and vice versa.
- ▶ Medicaid has been implemented state by state, but its **effectiveness varies by county**.

Spatial Machine Learning

Before COVID-19 (2019)

100% college2019 >= 0.5 0.4 42% hispanic2019 < 0.1 0.21 38% 0.32 19% 0.34 31% college2019 >= 0.63 hispanic2019 < 0.031 hispanic2019 < 0.13 medicaid < 0.8 0.51 9% 0.3 19% obesity2019 < 0.36 hispanic2019 < 0.04 food2019 >= 0.68 0.53 pop_2019 >= 380e-6 25 0.29 9% 0.3 0.44 0.27 14% 0.37 0.4 0.31 1% 0.72 1% 0.69 3% 0.2 0.51

Spatial Random Forest



MSE: 0.0168 / RMSE: 0.1296

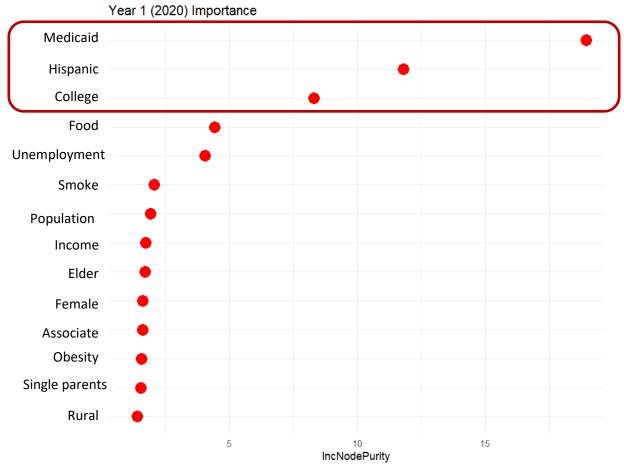
MSE: 0.0098 / RMSE: 0.0.0994

Spatial Machine Learning

Year 2 Effect (2021)

0.3 100% yes medicaid = 1 no -college2021 >= 0.55hispanic2021 < 0.17 0.37 0.29 unemploy2021 >= 0.18 college2021 >= 0.61 college2021 >= 0.53 food2021 >= 0.82 unemploy2021 >= 0.15 food2021 >= 0.85 hispanic2021 < 0.043 0.37 0.38 0.29 6% 0.25 16% 0.32 8% 0.48 7% 0.53 0.66 5%

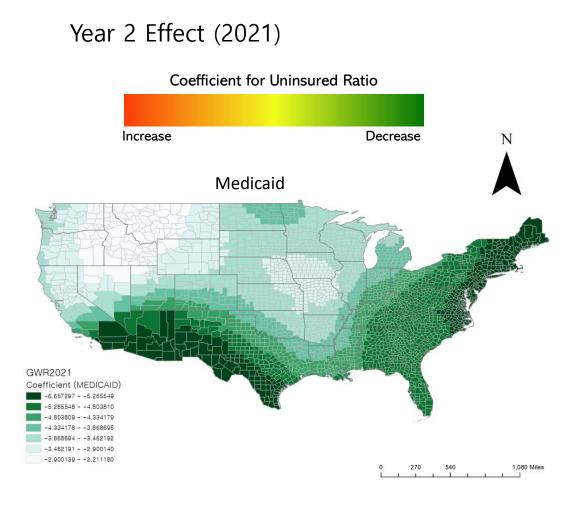
Spatial Random Forest

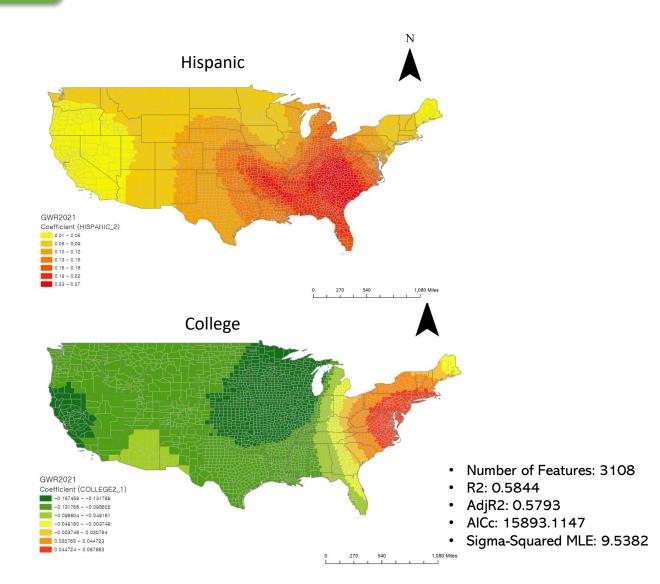


MSE: 0.0118 / RMSE: 0.1086

MSE: 0.0078 / RMSE: 0.0888

Geographically Weighted Regression





Conclusion

Conclusion

- ▶ This study reflected the spatial pattern of uninsured rates that occurred during COVID-19.
- Using Gi* Hotspot analysis, patterns were found at the county level.
- Spatial ML was used to calculate the importance of each variable for predicting the non-insurance rate.
- Coefficient values of important variables were examined by county through GWR.
- ▶ The results support previous studies that Medicaid affects the reduction of non-insured rates.
- Through Spatial ML, we found that Medicaid, Hispanic ratio, and education level were important variables during the COVID-19 period.
- GWR showed that the influence of each variable had spatial characteristics.
- ► The influence of Medicaid expansion policy affects each other by county's spatial characteristics and neighboring counties and interacts with other social and economic factors.

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Q & A

