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Feasibility of Generating Community Health Measures Using Geospatially-enabled Electronic Health Records

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Aim

Location data stored in electronic health records (EHRs) provide the opportunity to identify health disparities with geographic specificity and to target public health interventions to the most at-risk communities. Our aim is to use geospatially-enabled EHRs to develop and validate community health measures at geographic levels smaller than county.

Methods

Identification and Prioritization of Potential Community Health Measures

- We compiled a preliminary list of potential community measures using indicator definitions similar to HEDIS (Healthcare Effectiveness Data and Information Set). HEDIS measures are widely used by health care providers for performance measurement. However, they typically are not reported by geographic area.
- Public health stakeholders at the Marion County Public Health Department appended the list of potential to include other desired measures.
- We measured the perceived usefulness of the potential measures via a survey of Indiana public health professionals.

Feasibility Assessment

We evaluated the feasibility of generating each potential community measure based on:

- 1. Likelihood of electronic capture in an EHR system,
- 2. Availability within an information system accessible to public health,
- 3. Percent health care providers contributing data (indicating potential to represent population health based on proportion of health care providers contributing related data),
- 4. Disease prevalence (indicating potential statistical stability at small geographic scales) , and
- 5. Highest degree of geographic granularity available (enabling use at small scales).

Evaluation of the Representativeness of the EHR Data

 For a three-year period (2011-2013), we identified all unique patients in Marion County, Indiana whose records were captured in the Indiana Network for Patient Care (INPC). We calculated and mapped the ratio of unique INPC patients to total 2010 census population at the census tract level.

Results

We received survey responses from a cross-section of professionals in the public health field (See: *Figure 1*: *Organization Type of Respondents* and *Table 1*: *Public Health Role(s) of Respondents*).

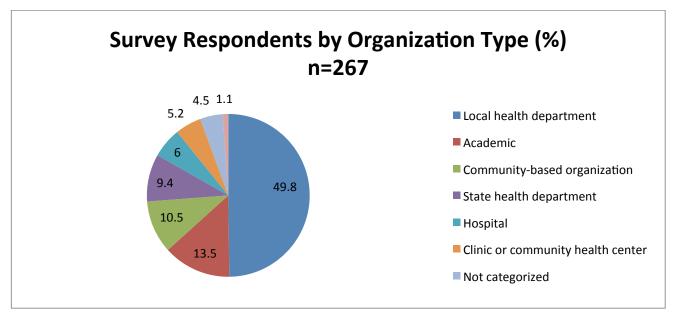


Figure 1: Organization Type of Survey Respondents

As shown in Table 1, the most common public health roles of survey respondents were health communication/education and senior administration. Environmental health and epidemiology/data analysis were also well represented roles.

Percent range	Public Health Role (% Respondents), n=209		
00.400/	Health communication/education (36.8%)		
30-40%	Senior administration/executive (33%)		
45.050/	Environmental health (22.5%)		
15-25%	Epidemiological analysis/data analysis/statistics (18.2%)		
	Community benefit (10%)		
	Public health nursing (9.1%)		
7-10%	Vital records (7.2%)		
	Management (7.2%)		
	Vital records (7.2%)		
<5%	Social work (4.3%), Research (3.8%), Marketing (3.3%), Admin support (2.9%), Med Technology (1.9%), Physician (1.4%), Geographic information analysis (1.4%)		

Table 1: Public Health Role(s) of Survey Respondents

Prioritization of Potential Community Health Measures

One hundred eighty six (186) respondents scaled the potential community health measures from "Very useful" to "Not at all useful". Associated scores were applied as follows: Very useful = 4; Somewhat useful = 2; Not very useful = 1; Not at all useful = 0

Based on the average score of each measure, we prioritized the preliminary list of twenty-four (24) potential community health measures into four tiers (see **Table 2**: Prioritization of Potential Community Health Measures based on Average Usefulness Score).

Average Usefulness Score n=209	Potential Community Health Measure		
Tier 1: Most Useful 2.4 – 2.6	 Prevalence of diabetes Prevalence of hypertension and other common cardiovascular diseases Prevalence of substance abuse Prevalence of asthma and chronic obstructive pulmonary disease Vaccination coverage for school age children Flu vaccination coverage 		
Tier 2: More Useful 2.0 – 2.3	 Hemoglobin A1c testing for patients with diabetes Cholesterol screening for patients with cardiovascular conditions Breast cancer screening Incidence of various cancers Prevalence of depression Incidence of chlamydia, gonorrhea, and syphilis HIV screening Prevalence of hepatitis B and/or hepatitis C HPV vaccination coverage Evidence of violence/trauma (e.g., domestic violence) 		
Tier 3: Less Useful 1.5 – 1.9	 Hemoglobin A1c controlled at <8% for patients with diabetes Emergency room utilization for people with asthma Cholesterol levels < 100 mg/dL for patients with cardiovascular conditions Chlamydia screening 		
Tier 4: Least Useful 1.3 – 1.4	 Prevalence of asthma among those with attention deficit hyperactivity disorder / impact of co-morbidity on emergency department visits Emergency room utilization by people with dental pain/infections Prevalence of dental caries 		

Table 2: Prioritization of Potential Community Health Measures based on Average Usefulness Score

Some respondents (n=69) identified additional community measures needed by public health professionals. Most of these were in the categories of built environment/community resources, access to care, health behaviors, health outcomes, or mental health. Measures of socio-economic condition, maternal child health, and obesity were also suggested.

While seventy-five percent (75%, 201/269) reported using or working with community health data, sixty-three percent (63%) of respondents either do not have access to <u>sub-county measures</u> of community health (99/211) or were unsure of whether they had access to such measures (33/211). Of those reporting access to and use of sub-county statistics (n=69), the most frequent uses were community health needs assessment, targeting of interventions, identification of high-risk groups, and identification of disparities. See: **Table 3**: Reported Use of Sub-County Measures.

Percent range	Reported Use of Sub-County Data n=69		
55-70%	For community health needs assessment (65.7%) To target interventions to appropriate populations (64.3%) To identify high-risk groups (61.4%) To identify disparities (58.6%)		
40-50%	For health improvement planning (47.1%) For program evaluation (42.9%)		
25-30%	25-30% For improvement of routine public health functions (28.6%)		
<6%	<6% Do not use available sub-county data (5.7%)		

Table 3: Reported Use of Sub-County Data

Feasibility of Community Health Measures

We deemed all fourteen of the proposed HEDIS-based measures and six of ten additional measures requested by our public health partners as feasible (See **Table 4**: Feasibility Matrix, on following page). All proposed measures met the criteria for geographic granularity, with geographic (X,Y) coordinates and block IDs available. These spatial attributes are generated on a daily basis for all clinical records entering and stored in INPC using a secure geocoding service that uses street address as the input location data. The more limiting criteria were electronic capture and percentage of HCPs contributing related data.

Proposed Community Health Measure	Feasibility Decision	1. Likelihood of Electronic Capture	2. System Accessibility	3. % Providers Contributing Data	4. Geographic granularity	5. Prevalence
Tier 1 - Most Useful	Based on	10 (very likely) – 1 (very	10 (very likely) –	% of INPC	10 (very granular) – 1 (no	per 1000, in
Prevalence of diabetes*	Items 1-5 Yes	unlikely) 10	1 (very unlikely) 9	providers 85%	granularity) 10	Marion County, IN 93
Prevalence of hypertension and other common cardiovascular diseases*	Yes	10	9	85%	10	319
Prevalence of substance abuse	Maybe	7	5	60%	10	
Prevalence of asthma and COPD*	Yes	10	9	85%	10	91
Vaccination coverage for school age children	Maybe	7	5	60%	10	
Flu vaccination coverage	No	7	5	30%	10	341

10

10

Yes

Yes

Tier 2 – More Useful

Prevalence of depression*

Prevalence of chlamydia, gonorrhea, and syphilis*

Prevalence of various cancers*

Breast cancer screening*

Colorectal cancer screening*

HbA1c testing for patients with diabetes*

LDL-C screening for patients with cardiovascular conditions* HIV screening

HPV vaccination coverage

Prevalence of viral hepatitis – HBV and (especially) HCV Evidence of violence/trauma (e.g., domestic violence)

Tier 3 – Less Useful

LDL-C Levels < 100 mg/dL for patients with cardiovascular conditions* Emergency room utilization for people with asthma* HbA1c controlled at <8% for patients with diabetes * Chlamydia screening *

Tier 4 - Least Useful

Prevalence of dental caries

Prevalence of asthma among those <u>with ADHD;</u> impact on ED visits Emergency room utilization by people with dental pain/infections **HEDIS measure*

7 Yes 10 85% 10 49.1 Yes 10 7 75% 10 721.8 5 Yes 10 50% 10 565.2 Yes 10 10 95% 10 9 Yes 10 95% 10 Yes 10 10 90% 10 7 3 10 97.2(M)/384.3(F) No 10% Yes 10 10 95% 10 11.5 3 3 No 60% 11.5

85%

90%

10

10

138

10.7, 3.5, 0.2

9

10

Yes	10	9	95%	10	
Yes	10	10	95%	10	48.8
Yes	10	10	95%	10	
Yes	10	10	90%	10	

Maybe	10	7	40%	10	
Maybe	5	6	85%	10	
Yes	7	7	95%	10	

Table 4: Feasibility Matrix

Validation of the Representativeness of the EHR Patient Population Using Spatial Attributes

The ratio of unique EHR patients to 2010 total population is not consistent across Marion County censu but rather demonstrates a geographic pattern, with the lowest ratios occurring in the northern portion of and the highest ratios in the central part of the county. See *Figure 2*: *Ratio of INPC Patients to Total P for Marion County Census Tracts*.

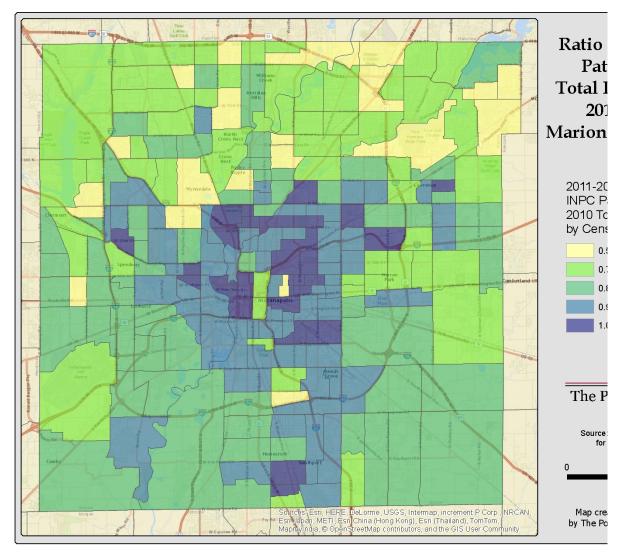


Figure 2: Ratio of INPC Patients to Total Population for Marion County Census Tracts (this map to be depict as ratios versus % overcount)

Discussion

Of the six measures determined to be the most useful to the public health community, we deemed three to be very likely feasible:

- 1. prevalence of diabetes,
- 2. prevalence of hypertension and other common cardiovascular diseases, and
- 3. prevalence of substance abuse.

EHRs appear to be geographically-biased. Further investigation is needed to determine the cause for this. Potential explanations for the apparent bias include:

- 1. Areas where residents have relatively short housing tenures will appear to have more people, since the patient counts from clinical data include data from spans of time during which several families might move in and out of a single address.
- 2. Lower patient counts may tend to occur in areas where residents are more likely to use healthcare providers whose data is less completely captured in an EHR.
- 3. In our study area, the EHR system has good capture of data from systems serving the uninsured and people with Medicaid, but has less complete outpatient data from those with private insurance.
- 4. Some residential addresses are associated with a central delivery location among scattered residences, as may happen with trailer parks and large apartment complexes, and as such can be incorrectly assigned to census geographies.

Next Steps

- Investigation of possible factors not related to health that are affecting patient counts.
- Exploration of whether the same pattern of patient ratios exist at other levels of geographic specificity, such as ZIP code and political boundaries.
- Development of adjustment factors to account for non-health factors affecting patient counts.
- Prototype measure development of the feasible, higher priority measures.

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Project Partners







The Polis Center





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WHAT

Use of electronic health records to develop community health measures with geographic specificity.

Identified Health Measures

|--|

Compiled list of potential measures based on definitions similar to HEDIS (Healthcare Effectiveness Data and Information Set) performance measures



Engaged public health partners to identify other measures of interest



Surveyed Indiana public health professionals to rank usefulness of potential measures

What Survey Respondents Told Us

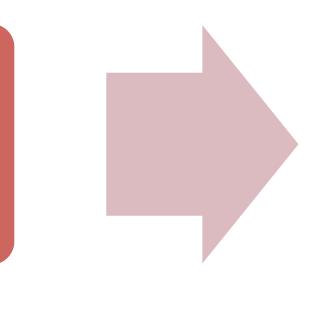
63% do not have access to sub-county measures or were unsure if they did. For those with access , the top reported uses were:

- Community health needs assessment (65.7%)
- To target interventions to appropriate populations (64.3%)
- To identify high-risk groups (61.4%)
- To identify disparities (58.6%)

Of our list of 24 potential EHR-based measures, those shown in **Table 1** would be the most useful.

Who Responded **Survey Respondents by Organization Type (%)** n=267 1.1Local Health Department 4.5 Academic Community-based Organization **State Health Department** 49.8 Hospital 10.5 Clinic or Community Health Center Not Categorized 13.5 Healthcare

Feasibility of Using Geocoded Electronic Health Records for Community Health Assessment



Assessed Feasibility

Assessed each potential, EHR-based community health measure on the following criteria:



✓ Disease prevalence

Geographic granularity

What Our Feasibility Assessment Told Us

Three of the top six measures ranked as the most useful by survey respondents are very likely feasible.

Table 1: Potential Measures Ranke

Prevalence of diabetes Prevalence of hypertension and othe diseases

Prevalence of substance abuse Prevalence of asthma and COPD Vaccination coverage for school age

Flu vaccination coverage

Next Steps

- Investigate possible non-health factors affecting EHR patient counts.
- code and political boundaries.
- patient counts.
- Prototype the higher priority community health measures that we deemed feasible.

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WHY

- Likelihood of electronic capture
- Accessibility of electronic record
- Percent health care providers contributing

ed the Most Useful	Feasible
	Yes
er common cardiovascular	
	Yes
	Maybe
	Yes
e children	Maybe
	No

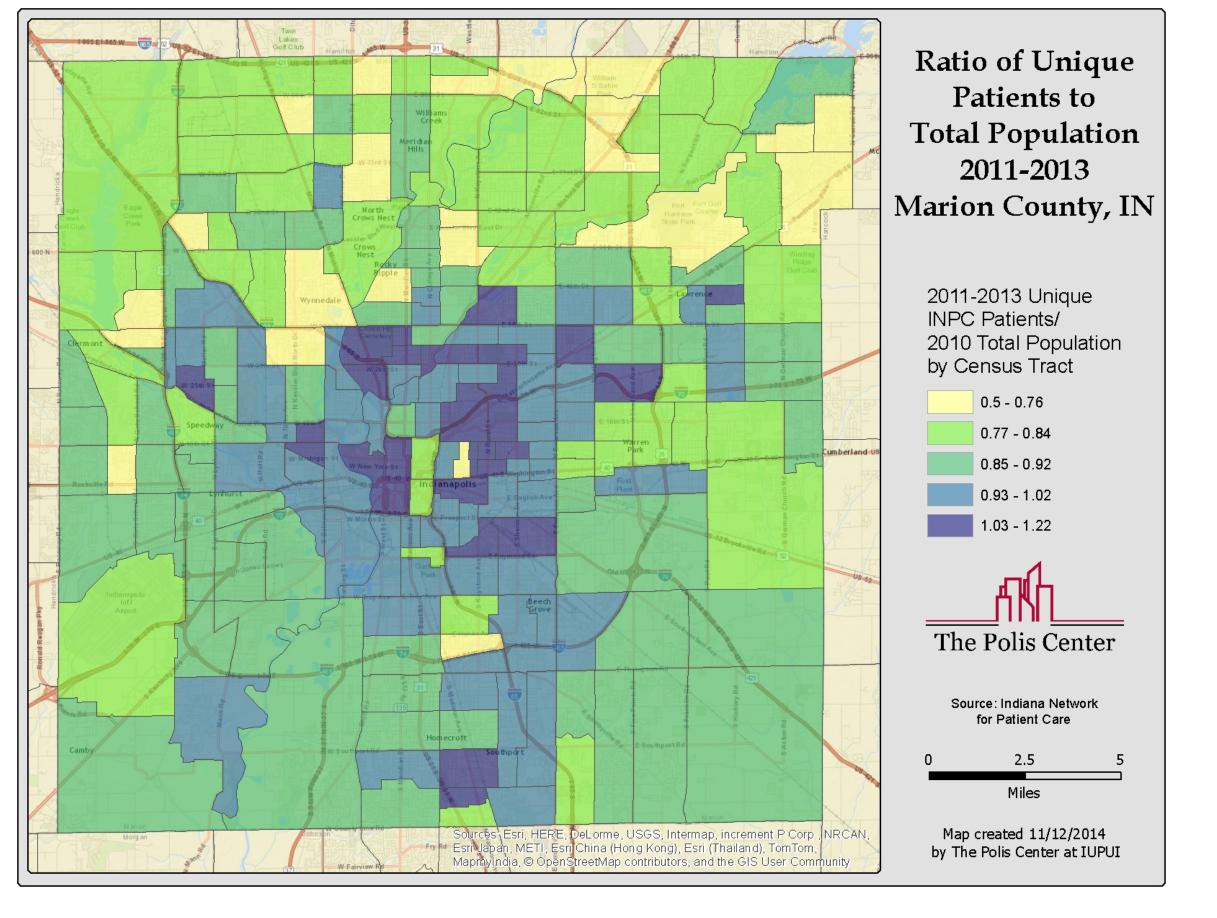
• Explore patient ratios at other levels of geographic specificity, such as ZIP

• Develop adjustment factors to account for non-health factors affecting



What Our Data Evaluation Told Us

The ratio of unique EHR patients to total census population is not consistent across the county. Rather, there is a geographic pattern, with lower ratios in the northern part of the county and higher ratios in the central part.



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To identify vulnerable populations and target public health interventions to the most at-risk communities.

Evaluated Data Representativeness

Identified all unique patients in each census tract for a 3- year time period

Calculated and mapped the ratio of unique patients to total population for each census tract using US census data







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