**Breakout Session 6: Track A** 

Consideration of Geospatial Distribution in the Measurement of Study Cohort Representativeness and Data Quality

> Dr. Keith Feldman (Moderator) Assistant Professor, Children's Mercy Kansas City

#### Contextualizing and Addressing Population-Level Bias in a Social Epigenomics Study of Asthma Supplement R01MD015409-03S1



Speaker: Keith Feldman, PhD, PI: Elin Grundberg PhD





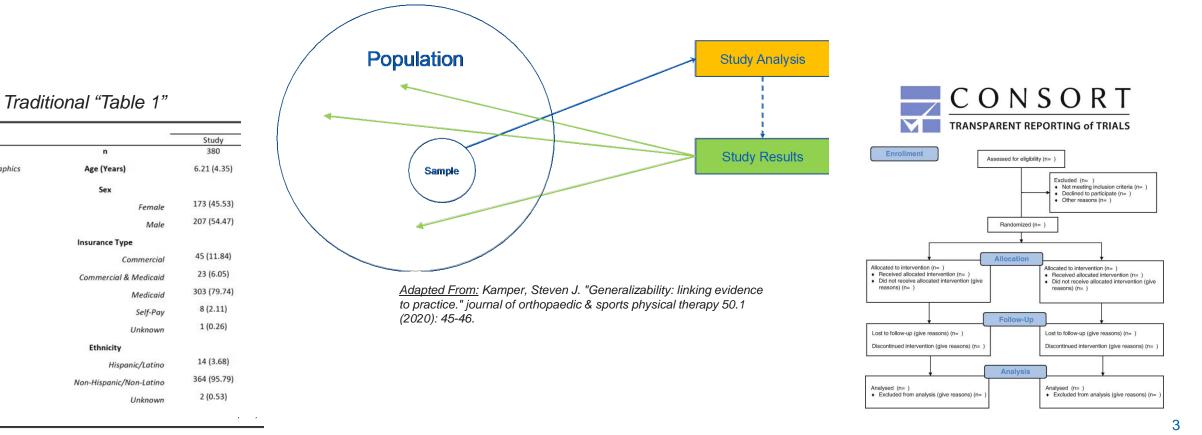


The Problem with Generalization

Highlights

Study results are a product of the data (subjects) used for analysis

Challenges



**Future Work** 

LOVE WILL.

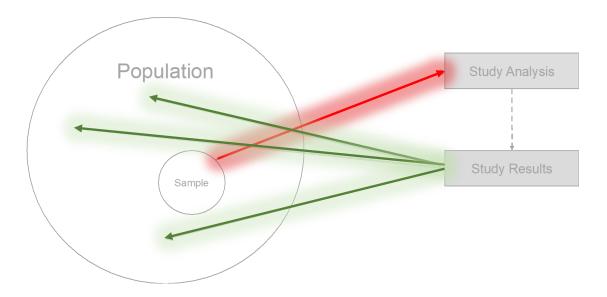
Demographics

Introduction





Develop methods to quantify the representativeness of a study sample against a reference cohort to better understand for whom the study results can be expected to generalize, and to develop methods robust to such imbalances







**Future Work** 

# Quantify Representativeness

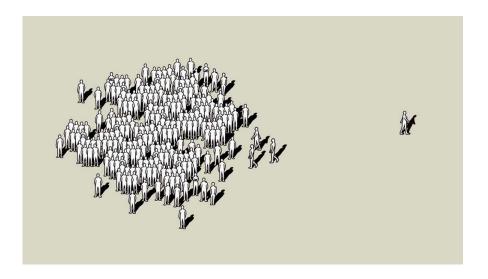
#### **Geospatial Representation**

Benchmarking characteristics of study cohort against a reference population on whom you want to assess potential generalizability.





Quantify patterns of data quality across subgroups of data and over time. Assess impact of data quality issues on reliability / stability of study results





## Parent Award Cohort and Data

The Stress, Epigenome and Asthma (SEA) study focuses on disentangling mechanisms by which exposure to chronic stress may cause epigenetic changes increasing susceptibility to rhinoviral (RV) infection – and thus increased risk of asthma exacerbations

- Study Cohort
  - SEA study (03/15/2021 -11/08/2022) 400 English-speaking Black/African American children, age <18 presenting to ER with acute respiratory symptoms related to an asthma diagnosis
- Reference Cohort (same period)
  - 12,699 Black/African American children with a history of Asthma.
  - 2,757 children were identified as having at least 1 encounter for an Asthma exacerbation during the study period.

#### Location:

Distance between the individual's address and recruitment location (CMKC Adelle Hall)

#### **Socio-Demographics:**

Age at encounter, primary insurance, sex assigned at birth, as well as self-reported race, ethnicity, and nationality.

#### **Clinical**:

Pediatric comorbidity index for all visits during the study period and derived count of eligible visits during the study.

Population-Level Indicator Data:
 Fraction of population >25 with educational attainment of at least high school graduation, and fraction of houses that are vacant.



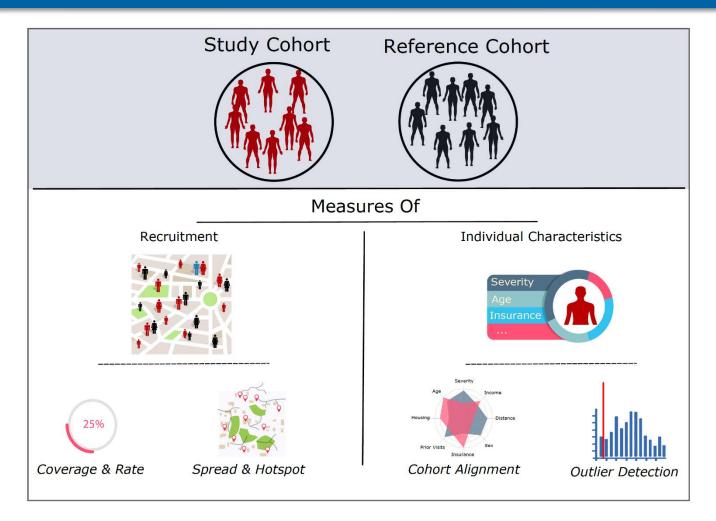


Introduction

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Future Work

## Geospatial Assessment





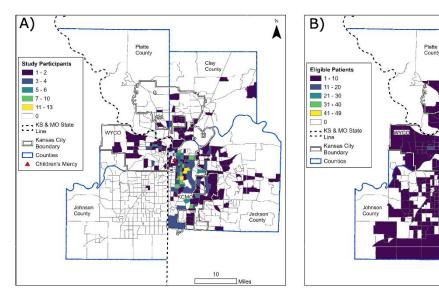


Future Work

## Measures of Recruitment

Coverage and recruitment	Coverage: Number of distinct regions of recruited subjects are drawn from compared to number of possible regions for recruitment.Can be calculated at any granularity of geospatial geography 
Spread and Hotspots	<u>Spread:</u> The total geographic area across the unique geographies that comprise the study cohort. To identify situations in which high-recruitment represents a limited geographic area. <u>Hot Spots:</u> Measure of geographic clustering to identify an area with a higher concentration of recruited subjects compared to the expected number given a random distribution of subjects.

**Recruitment Hot Spots:** Results of the Getis-Ord Gi\* analysis on SEA recruitment by census tract. Results indicate census tracts in which a higher-than-expected count of SEA study participants were enrolled relative to nearby census tracts



Coverage of approximately 39.1% (149 Tracks Study Cohort, 380 Tracks Reference).



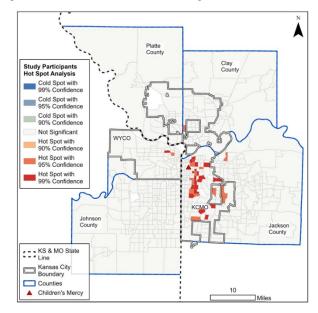


Future Work

#### Measures of Recruitment

Coverage and recruitment	Coverage: Number of distinct regions of recruited subjects are drawn from compared to number of possible regions for recruitment.Can be calculated at any granularity of geospatial geography (Counties, Census Block groups, etc.) as defined by the user.Recruitment rate: geographic area that are recruited to the study.			
	This value can then be aggregated across the set of geographic areas in the reference set to summarize recruitment averages and variability.			
Spread and Hotspots	<u>Spread:</u> The total geographic area across the unique geographies that comprise the study cohort. To identify situations in which high-recruitment represents a limited geographic area.			
	Hot Spots: Measure of geographic clustering to identify an area with a higher concentration of recruited subjects compared to the expected number given a random distribution of subjects.			

**Recruitment Distribution:** Count of patients from the SEA study cohort (Panel A) and reference cohort (Panel B) within each of the 2020 US Census Tracts for the catchment area of Children's Mercy Kansas City



**Spread:** Although a moderately sized absolute number, the 149 unique tracts represent a relatively small geographic area of just 234.47 sq. miles. This represents only about 23% of the total coverage of the total area reference cohort (1000.27 sq. miles).





#### Measures of Recruitment

Case Control	<ul> <li>Create an aggregate measure of how well aligned the study and reference cohorts are given a set of factors.</li> <li>Measures deviation between the study and reference populations as "balance" in case-control matches.</li> <li>For each study cohort members, 1 random individual is drawn from the reference cohort in the same geographic area.</li> <li>Pairs are treated as "matches" and can directly leverage an array of established case-control balance metrics, including measures of standardized mean difference (continuous), and maximal proportion difference (nominal) data.</li> </ul>
Distance Based	Identify Individual-level Measures of Similarity + Outlier Detection For each member of the study cohort, distance is computed against all individuals from of the reference population drawn who resides in the same geographic region. Used to compute a measure of how likely the average distance between study member and reference cohort is to happen by chance. Used to compute a measure of the expected intra-individual distance within the reference cohort. How close is the distance between study member and average reference distance.

**Balance of study cohort and reference population:** 500-bootstrap iterations of matching. Continuous variables are summarized with a Standardized Mean Difference (SMD), while nominal factors capture the largest proportion difference within levels of the respective factor. Panel (A) represents matching performed within the matching geography of each study patient.

	Variable	Mean (SD)	[Min-Max]
(A) - Within Geography	Age	0.471 (0.294)	[0.003-1.651]
	Sex	0.049 (0.038)	[0-0.194]
	Insurance Type	0.058 (0.034)	[0.006-0.158]
	Ethnicity	0.036 (0.008)	[0.006-0.059]
	Acute Care Visits in Study Period	0.366 (0.093)	[0.047-0.6]
	Pediatric Comorbidity Index	0.52 (0.108)	[0.236-0.813]
	Distance to CMKC	0.018 (0.012)	[0-0.062]



## Measures of Recruitment

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**Overview Distance-B**ased Outliers: This table outlines each of the 6 identified outliers in the measures of individual characteristics. For each case, checkmarks in row (P) specifies in the data for the outlier, while (R) provides the distribution of each factor in the reference population

			Sex		Ethnicity		Insurance Type				
		Age	Female	Male	Hispanic	Non-Hispanic/ Non-Latino	Commercial	Commercial & Medicaid	Medicaid	Self- Pay	Acute Care Visits in Study Period
1	P	0.25		$\checkmark$		$\checkmark$	$\checkmark$				15.00
	R	9.40 (5.05)	5 (50.00)	5 (50.00)	0 (0.00)	10 (100.00)	1 (10.00)	3 (30.00)	6 (60.00)	0 (0.00)	3.00 (3.37)
	P	5.92	$\checkmark$			$\checkmark$			$\checkmark$		4.00
2	R	9.81 (6.03)	4 (44.44)	5 (55.56)	0 (0.00)	9 (100.00)	0 (0.00)	2 (22.22)	6 (66.67)	1 (11.11)	2.00 (0.87)
	P	0.92		$\checkmark$		$\checkmark$	$\checkmark$				12.00
3	R	8.40 (5.41)	13 (52.00)	12 (48.00)	0 (0.00)	25 (100.00)	2 (8.00)	0 (0.00)	22 (88.00)	1 (4.00)	2.80 (2.12)
	P	3.17		$\checkmark$	$\checkmark$			$\checkmark$			6.00
4	R	7.92 (2.60)	2 (50.00)	2 (50.00)	0 (0.00)	4 (100.00)	2 (50.00)	0 (0)	2 (50.00)	0 (0.00)	2.00 (0.82)
	P	2.50		$\checkmark$		$\checkmark$			$\checkmark$		7.00
5	R	8.76 (4.78)	4 (23.53)	13 (76.47)	0 (0.00)	17 (100.00)	5 (29.41)	2 (11.76)	10 (58.82)	0 (0.00)	2.53 (1.55)
	P	5.00		$\checkmark$	$\sim$				$\checkmark$		6.00
6	R	6.42 (4.33)	12 (52.17)	11 (47.83)	2 (8.70)	21 (91.30)	2 (8.70)	0 (0.00)	21 (91.30)	0 (0.00)	2.87 (2.16)



Highlights

Challenges

Future Work

# Data Quality

Assume we have 3 variables: Age, Sex, Recruitment Location

Cour	Male			
Sex	Female			
Age	0-25 <sup>th</sup>	$25^{th} - 50^{th}$		
(percentile)	50 <sup>th</sup> – 75 <sup>th</sup>	75 <sup>th</sup> – 100 <sup>th</sup>		
Recruitment	Inpatient (1)			
Location	ED (2)			

#### Goal 1: Identify % of data in various subgroups

Configuration	Group_Val	Num_Params	Size	Val_AgeQ_1	Val_AgeQ_2	Val_AgeQ_3	Val_AgeQ_4
Female	(Sex,)	1	135	0.207407	0.244444	0.229630	0.318519
Male	(Sex,)	1	165	0.284848	0.254545	0.266667	0.193939
1	(Location,)	1	265	0.252830	0.264151	0.241509	0.241509
2	(Location,)	1	35	0.228571	0.142857	0.314286	0.314286
('Female', 1)	(Sex, Location)	2	120	0.216667	0.241667	0.225000	0.316667
('Female', 2)	(Sex, Location)	2	15	0.133333	0.266667	0.266667	0.333333
('Male', 1)	(Sex, Location)	2	145	0.282759	0.282759	0.255172	0.179310
('Male', 2)	(Sex, Location)	2	20	0.300000	0.050000	0.350000	0.300000

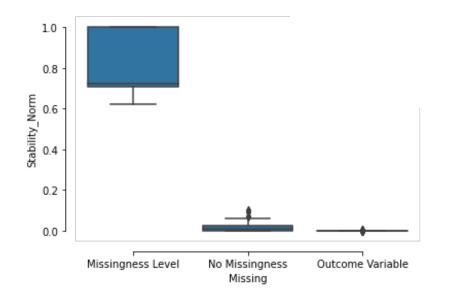




# Subgroup Bias Metric

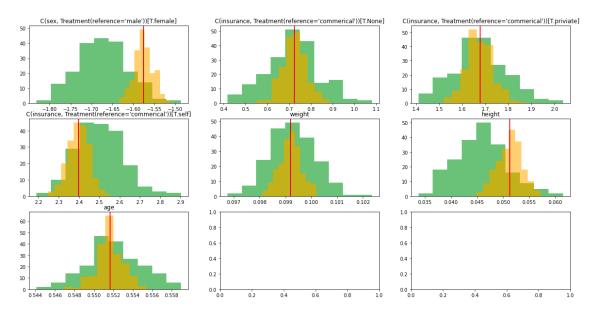
## Aggregate Measure

Developed a series of metrics to quantify the degree to which missingness for a given subspace significantly differs from levels of the factors that comprise it.



#### **Impact Analysis**

Verify if missingness in sub-groups results in biased estimates for downstream analysis, or if wider confidence intervals will account for this



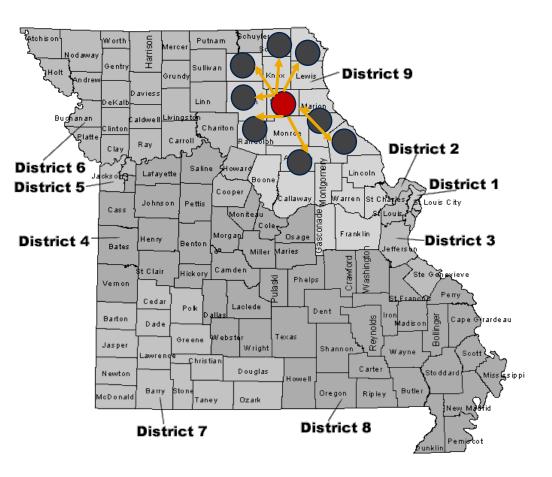


Challenges

**Future Work** 

## **Geospatial Boundaries**

- Comparing individuals within exactly matching geographic regions draws an arbitrary delineation in space, where subjects only a small absolute distance apart may be separated by a census block line and are thus not considered in the alignment measure.
- We introduce extensions of the previously described methods that leverage the entire reference population. This provides two benefits:
- •Allows for improved estimates of representations for study cohort in regions with small reference populations.
- •Second, it allows for the inclusion of data captured at an aggregate/population level (e.g., average access to mental Health Providers).





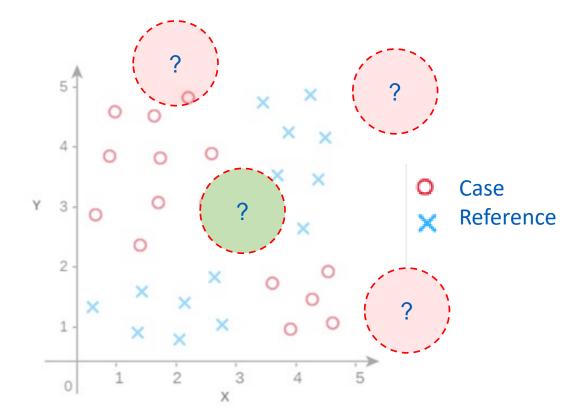
# Extrapolation vs Out of Reference

Highlights

Does the Data Represent The Full Range of Expected Values

Future Work

Challenges





Introduction

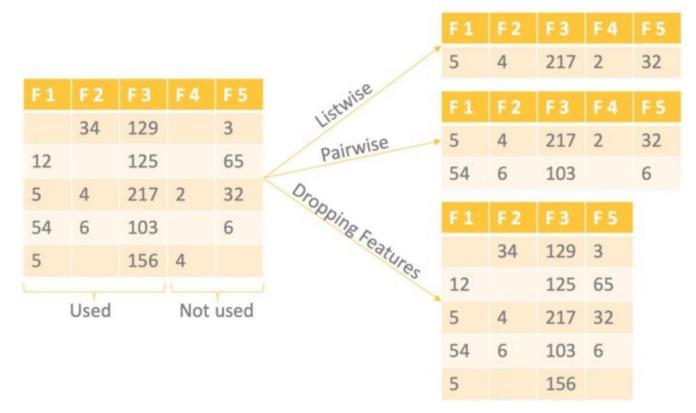


Highlights

#### In Real-World Studies, complete data of any kind is rare

Challenges

Future Work





Introduction

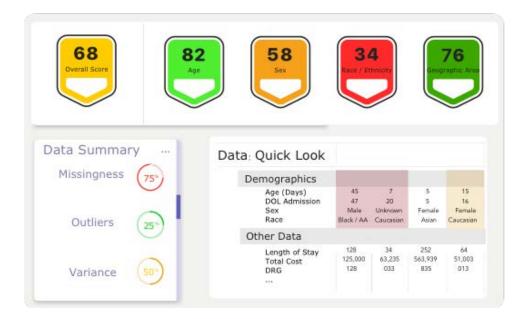


Challenges

Future Work

# Ongoing Work

Standardized metrics pertaining to the study cohort representativeness and quality will help inform researchers of inherent biases, limits of conclusions and more broadly improve generalizability for future cohort studies, allowing for safer reuse of data



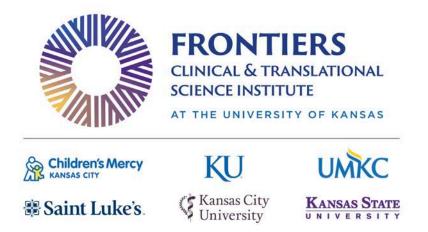




# Integration into CTSI Informatics Core

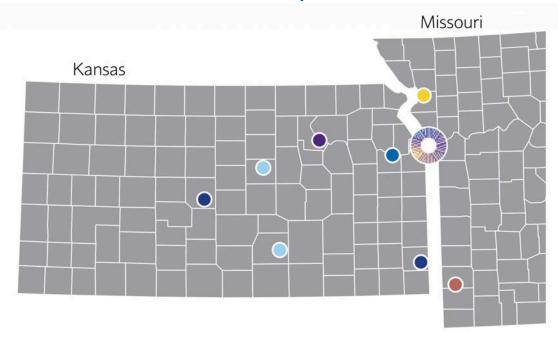
Highlights

Challenges



As part of the informatics core of our regional CTSI program, geospatial measurement tools are being created. Working with this core, led by CO-I Dr. Mark Hoffman, we hope to integrate these measures into the developed toolset for researchers

**Future Work** 



Introduction

