



# Getting the Most out of RWD through Enrichment of the Data

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# Agenda

RWD Definition for this talk

Limitations of RWD

Enrichment Definition

Examples

The Future





# Introduction

- RWD from EMR and Claims is driving significant value in Life Sciences to identify patient journeys, adverse events, market penetration patterns and potential identification of clinical trial candidates among other uses. This talk will focus on getting even more value out of RWD by enriching that data.

# What do we mean by Enrichment?

Extending data through linkage to other data sets where there are attributes that do not exist in the original RWD.

# Can we get more out of the data through enrichment?

Let's see a few examples.

# SDOH Example

## ADVANTAGES

- Be able to adjust the messaging, therapies and interventions to meet the needs of the individual.
- Engage underserved populations
- Be able to potentially broaden the clinical trial participant pool.
- HL7 FHIR Accelerators defining attributes

## ISSUES

- Need to have detailed zip code or geo-location data to connect the data
- Wide variance in attributes to consider (food availability tailored to the individual, financial data, health care services, cognitive ability),

# Genetics Example

## GENETIC TO DISEASE CONNECTION

Population Comorbidity Analysis						
Condition Term	Comorbidity Terms	Variant Disease Association	No. Comorbid	% Comorbid	No. Patients Universe	% Frequency (Universe)
Hypertrophic cardiomyopathy	Essential hypertension	Risk Factor	31,558	67.09%	8,868,477	15.67%
	Hyperlipidemia		24,519	52.34%	7,466,066	10.81%
	Congestive heart failure	Pathogenic	13,612	28.94%	1,205,105	2.15%
	Cardiomegaly	Pathogenic	12,864	27.35%	644,360	1.18%
	Atrial fibrillation	Pathogenic	12,514	26.60%	1,462,021	2.51%
	Cardiac arrhythmia	Pathogenic	12,295	26.14%	3,336,759	2.41%
	Anemia	Pathogenic	12,005	25.52%	3,777,923	4.67%
	Gastroesophageal reflux disease	Pathogenic	11,891	25.28%	4,275,874	6.11%
	Type 2 diabetes mellitus without complication		11,754	24.99%	3,229,169	5.90%
	Atherosclerosis of coronary artery without angina pectoris		11,632	24.73%	1,424,730	2.60%
	Gastroesophageal reflux disease without esophagitis		11,551	24.56%	2,858,011	5.22%
	Primary cardiomyopathy		10,402	22.11%	335,250	0.61%

  

Gene Disease Associations						
			Condition Term	Hypertrophic cardiomyopathy		Score
Gene Disease Association	Gene Name	Variant Disease Association	Gene Variant	Links	Score	Links Score
<input type="checkbox"/> Pathogenic	<input type="checkbox"/> MYH7	<input type="checkbox"/> Pathogenic	nsv513807	1	0.87	1 0.87
			rs1060501452	1	0.87	1 0.87
			rs121913624	1	0.87	1 0.87
			rs121913625	1	0.87	1 0.87
			rs121913626	1	0.87	1 0.87
			rs121913627	1	0.87	1 0.87

## DETAILS

- Many sources of data available to link a condition to a gene
- Need to standardize the condition data across a large variance in representations (ICD10, SNOMED, others),
- Once the gene is linked can extend to variants, proteins, biopathways

# Where does the value of genetic enrichment of Big Data come in?

- Identifying novel genetic relationships between conditions
- Estimating the prevalence of mutations
- Identifying disease occurrence of rare disease based on profiles of built on genetically related comorbidities that pre-date the disease of interest,
- Improving tertiary genetic analysis by identifying suspect uncertain variants that may well be pathogenic based on frequency of co-occurring comorbid conditions
- Identifying new targets for therapies based on related biopathways identified through comorbid genetic relations that point to new etiologies





# Summary

The enrichment of RWD has many benefits that go beyond the initial data use cases and is economically justified.

Gen AI is going to open up many other benefits of the enrichment of RWD.

# Thank You

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