

Concept Prevalence Design diagnostics

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Can we

Concept Prevalence

Network study: Conce the OHDSI Network

Researchers



aostropolets Anna Ostropolets

We want to announce a new netw https://github.com/OHDSI/StudyP

The full protocol can be found her https://github.com/OHDSI/StudyPudyProtocol_v0.1.docx 32

We want to study the usage patte useful information to answer many granularity of codes captured in a can be coded as ICD9 code 585.2 unspecified or even as 586 Renal Currently, researchers have no wa available for selection, or whether information to define the cohort or distribution of the concepts is limit Studies are dependent on cohort

In an ideal world, a cohort definition community. We would like to make

- Unique values in the *_con
- Unique values in the *_sou
- Mappings between them

Investigating Concept Heterogeneity and the OHDSI Network

Study Status Design Finalized

- Analytics use case(s): Characterization
- Study type: Methods Research
- Tags: -
- Study lead: Anna Ostropolets
- Study lead forums tag: aostropolets
- Study start date: March 15, 2019
- Study end date: -
- Protocol: Word file
- Publications: -
- Results explorer: -

This study aims to investigate the concept usa cohort identification to facilitate cross-instituti CDM, we will investigate if it is used at each sit how differences in corresponding source code

Counts of rea

No patient-

cou:

select condition_concept_id, count(*), 'condition'
from condition_occurrence
group by condition_concept_id

concept_id	cnt	type
201286	12478349	condition
41647883	26477	drug
2083274	623845	procedure

Additional information

In R, use the following commands to download and install:

install.packages("devtools") devtools::install_github("https://github.com/ohdsi-studies/ConceptPrevalence")

library('ConceptPrevalence')

Add inputs for the site:

As a side effect, we would also get a state and a side effect, we would also get a state and a side effect, we would also get a state and a side effect, we would also get a state and a side effect, we would also get a state and a side effect, we would also get a state and a side effect, we would also get a state and a side effect, we would also get a state and a side effect, we would also get a state and a side effect, we would also get a state and a side effect, we would also get a state and a side effect, we would also get a state and a side effect.

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Current results: 22 datasets from six countries





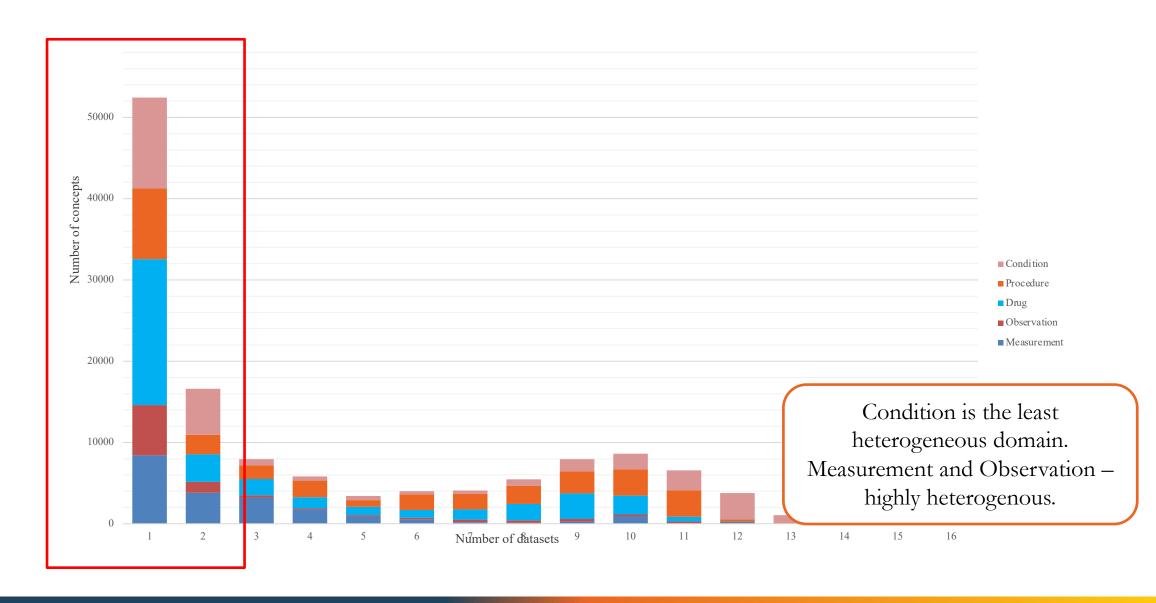
~375 billion records

14.6% of all databases within the OHDSI

> 1 million
distinct concepts

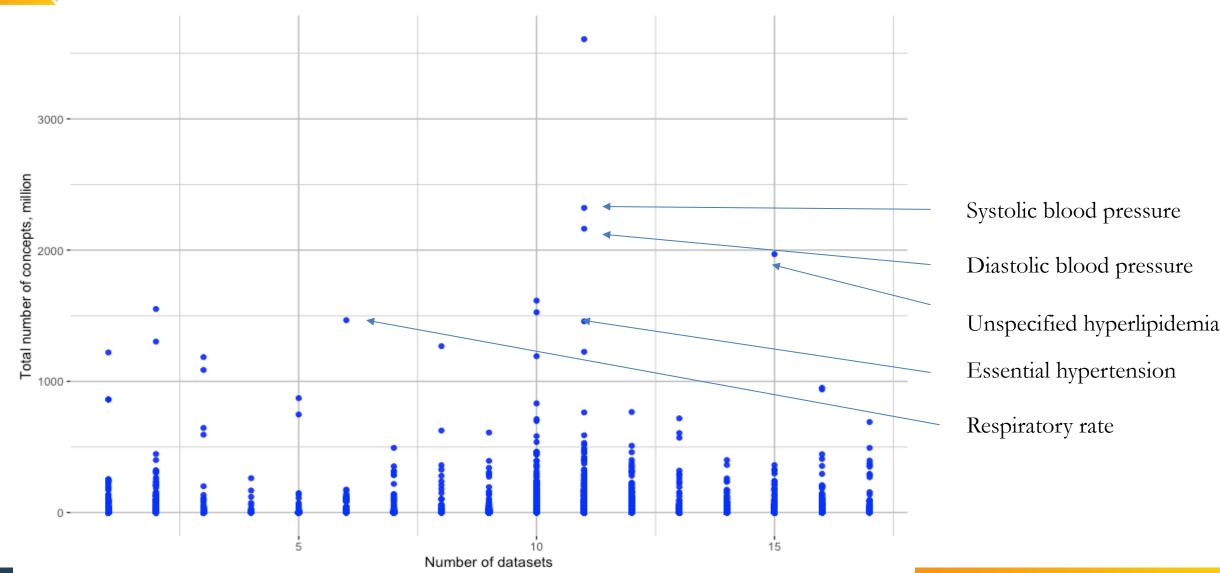


Most of the concepts can be found only in 1 dataset





Even common concepts do not appear in all datasets





Different databases have different granularity

DA France	JMDC	LPD AU	AmbEMR	CUMC	MDCD	Average					
		23%	4%	0.01%		2.2%	Database with broad concepts				
94%	90%	32%	17%	25%	13%	24.7%	concepts				
5%	9%	45%	69%	64%	68%	59.4%					
0.04%	1%		10%	9%	19%	13.1%	Database with specific				
							concepts				
0.04%	0.1%		0.2%	1%	1%	On avec	orago IIS sources are more				
							erage, US sources are more ar than international and US				
			0.01%				ns – more granular than EHR				
	94% 5% 0.04%	France JMDC 94% 90% 5% 9% 0.04% 1%	France 23% 94% 90% 32% 5% 9% 45% 0.04% 1%	France JMDC AU 4% 23% 4% 94% 90% 32% 17% 5% 9% 45% 69% 0.04% 1% 10%	France JMDC AU AmbEMR CUMC 23% 4% 0.01% 94% 90% 32% 17% 25% 5% 9% 45% 69% 64% 0.04% 1% 10% 9% 0.04% 0.1% 0.2% 1%	France JMDC AU Ambema CUMC MDCD 23% 4% 0.01% 94% 90% 32% 17% 25% 13% 5% 9% 45% 69% 64% 68% 0.04% 1% 10% 9% 19% 0.04% 0.1% 0.2% 1% 1%	France JMDC AU AmbEMR CUMC MDCD Average 94% 90% 32% 17% 25% 13% 24.7% 5% 9% 45% 69% 64% 68% 59.4% 0.04% 1% 10% 9% 19% 13.1% 0.04% 0.1% 0.2% 1% 1% On average				

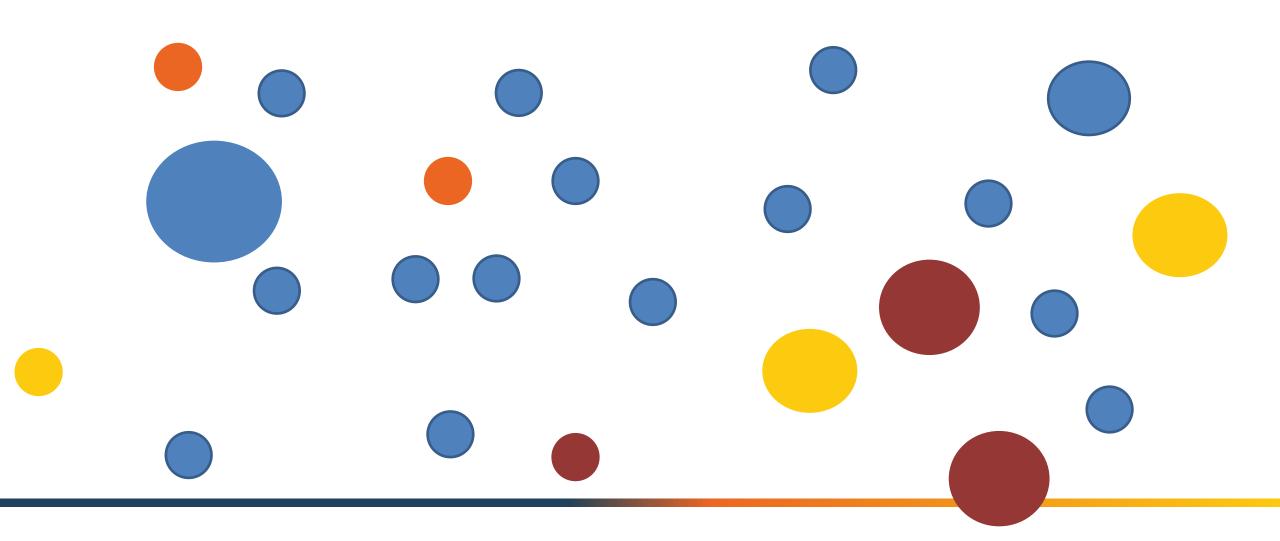


OHDSI's standardized vocabularies

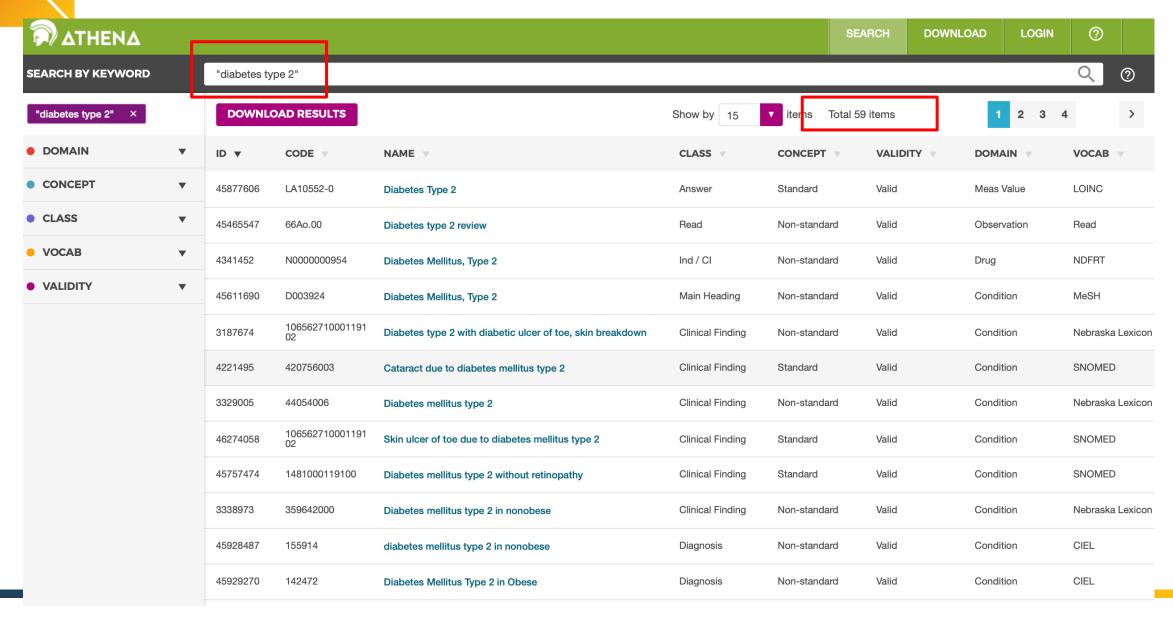
- 153 Vocabularies across 41 domains
 - MU3 standards: SNOMED, RxNorm, LOINC
 - Disparate sources: ICD9CM, ICD10(CM), Read, NDC, Gemscript, CPT4, HCPCS...
- >9 million concepts
 - ->3.3 million standard concepts
 - >5.1 million source codes
 - >629,000 classification concepts
- >55 million concept relationships
- >84 million ancestral relationships



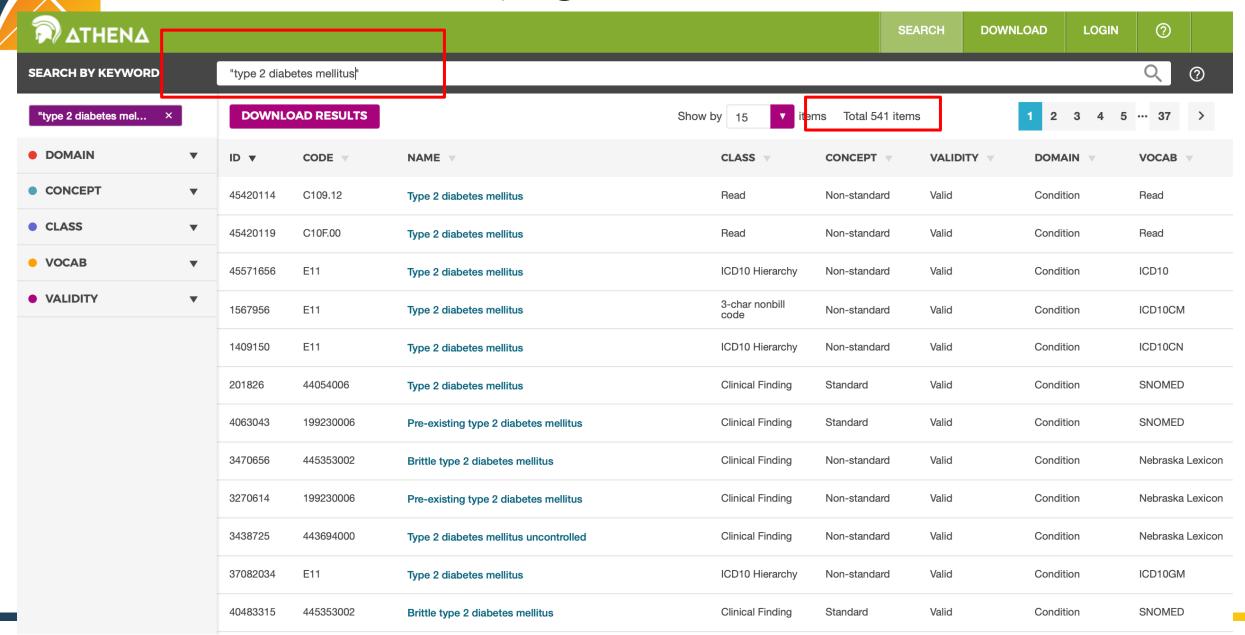
Given variety and heterogeneity of the data where do we start to create phenotypes?



Trying to select diabetes



Trying to select diabetes

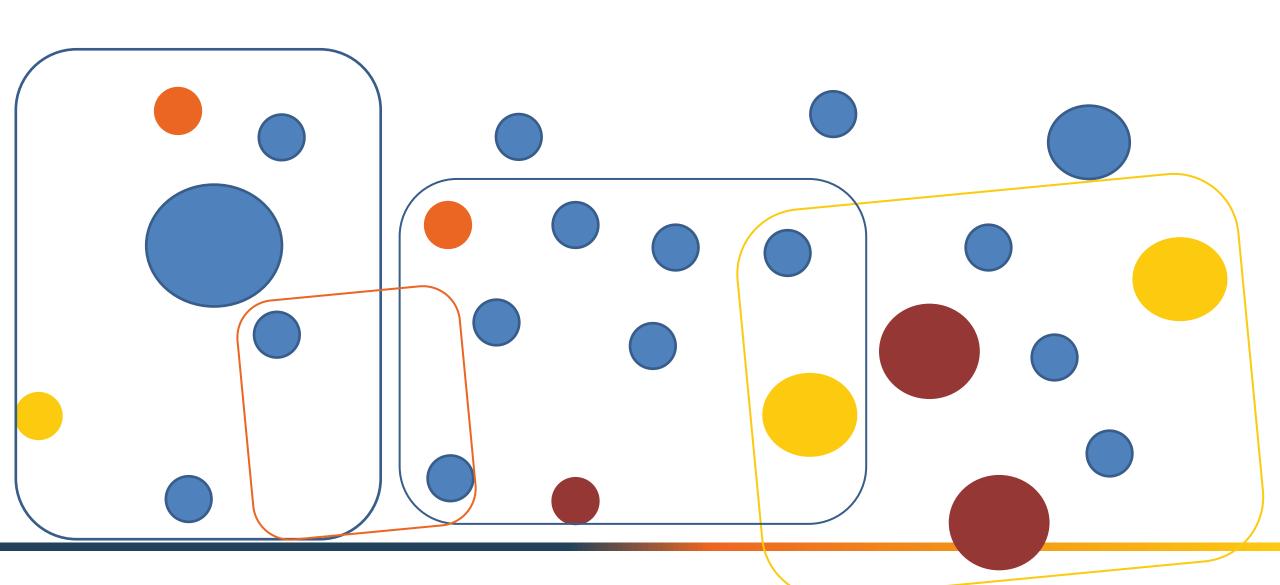


Trying to select diabetes

		<u> </u>									
Σ ΔΤΗΕΝΔ						SE	ARCH [OOWNLOAD	LOGIN	?	
SEARCH BY KEYWORD		"type II diab	etes"							Q	?
"type II diabetes" ×		DOWNLO	DAD RESULTS		Show by 15 v ite	ms Total 163 item	าร	1 2	3 4 5	11	>
DOMAIN	•	ID ▼	CODE ▼	NAME ▼	CLASS ▼	CONCEPT ▼	VALIDITY	DOMA	IN ▼	VOCAB	V
CONCEPT	•	45453110	C109.13	Type II diabetes mellitus	Read	Non-standard	Valid	Conditi	on	Read	
• CLASS	•	45486691	C10F.11	Type II diabetes mellitus	Read	Non-standard	Valid	Conditi	on	Read	
VOCAB	•	40386778	190384004	Type II diabetes mellitus	Clinical Finding	Non-standard	Invalid	Conditi	on	SNOMED	
• VALIDITY	•	3121806	190384004	Type II diabetes mellitus	Clinical Finding	Non-standard	Invalid	Conditi	on	Nebraska	Lexicon
		40482801	443694000	Type II diabetes mellitus uncontrolled	Clinical Finding	Standard	Valid	Conditi	on	SNOMED	
		45443104	C10FJ11	Insulin treated Type II diabetes mellitus	Read	Non-standard	Valid	Conditi	on	Read	
		45473337	C109J12	Insulin treated Type II diabetes mellitus	Read	Non-standard	Valid	Conditi	on	Read	
		45503184	C10P100	Type II diabetes mellitus in remission	Read	Non-standard	Valid	Conditi	on	Read	
		45463265	C10FG11	Type II diabetes mellitus with arthropathy	Read	Non-standard	Valid	Conditi	on	Read	
		45499864	C109G11	Type II diabetes mellitus with arthropathy	Read	Non-standard	Valid	Conditi	on	Read	
		45423316	C10F511	Type II diabetes mellitus with gangrene	Read	Non-standard	Valid	Conditi	on	Read	

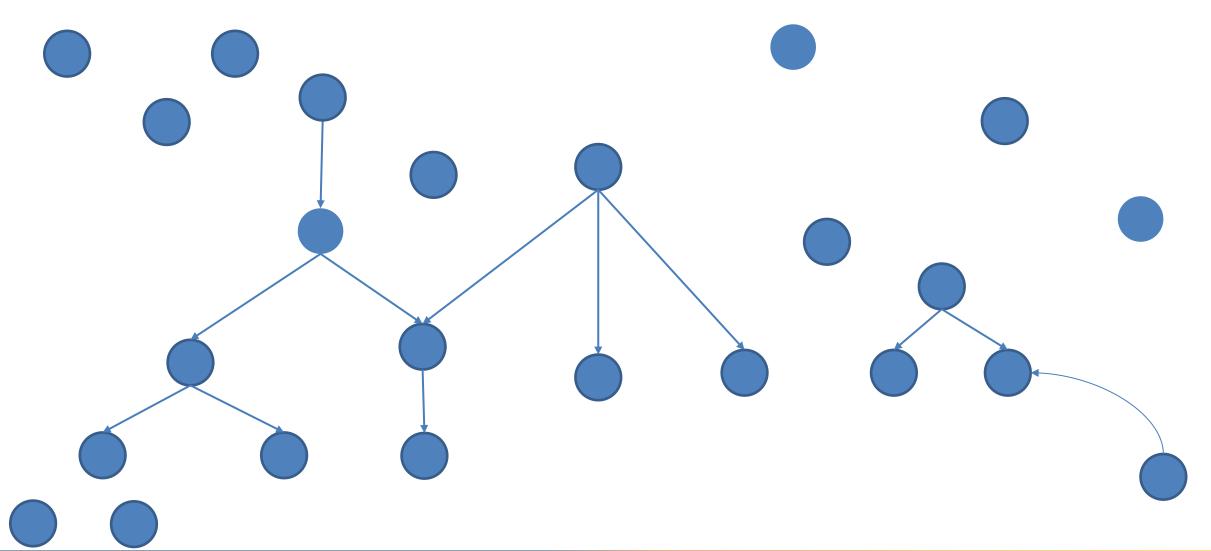


Even if we got the starting point, how do we select all the concepts that are relevant?



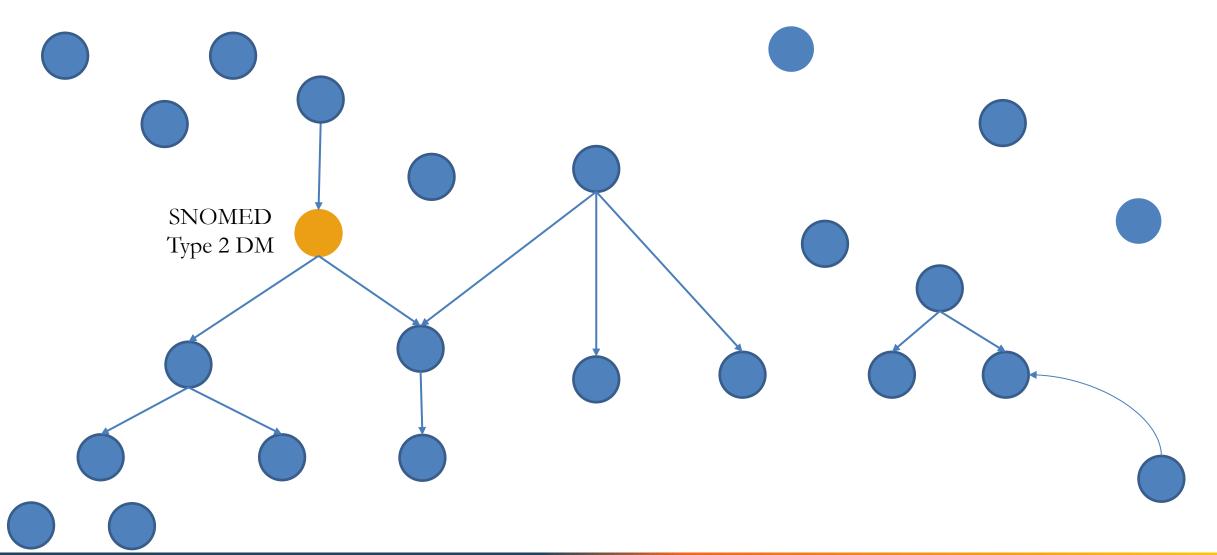


Select the starting point



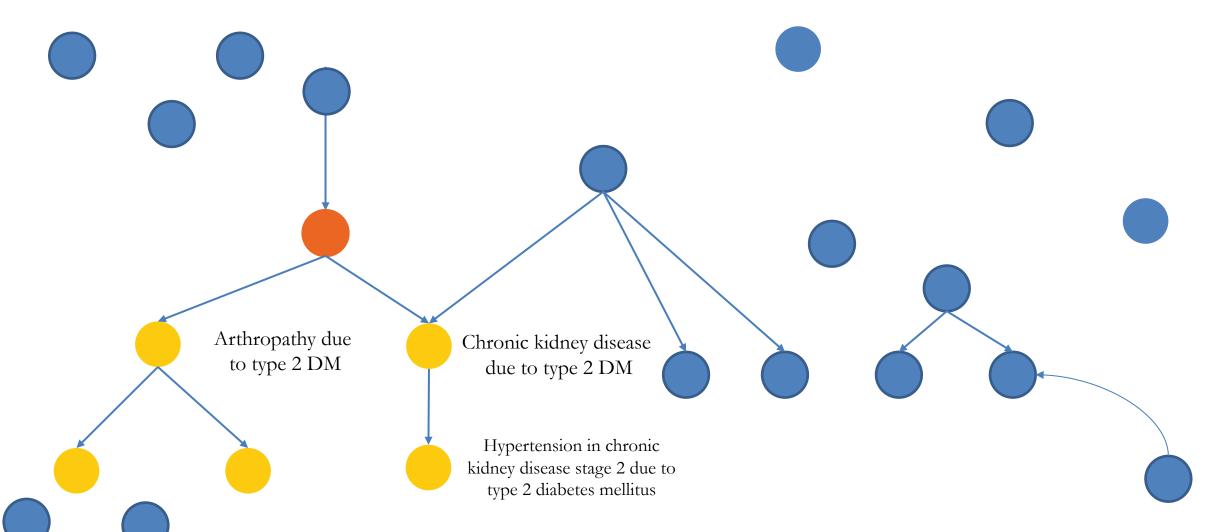


Select the starting point



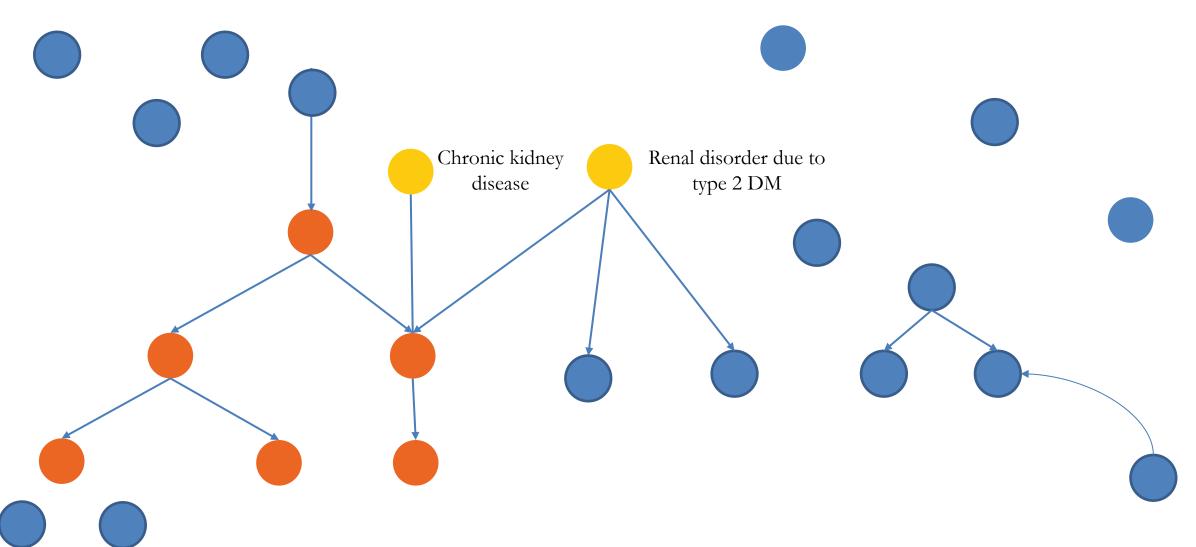


Select descendants



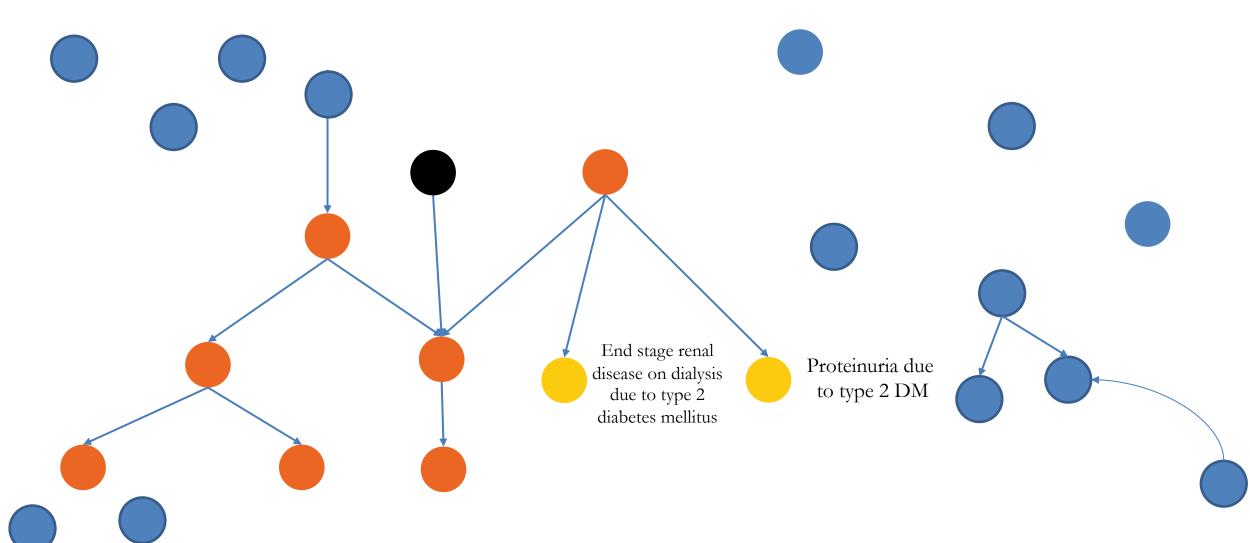


Check non selected parents



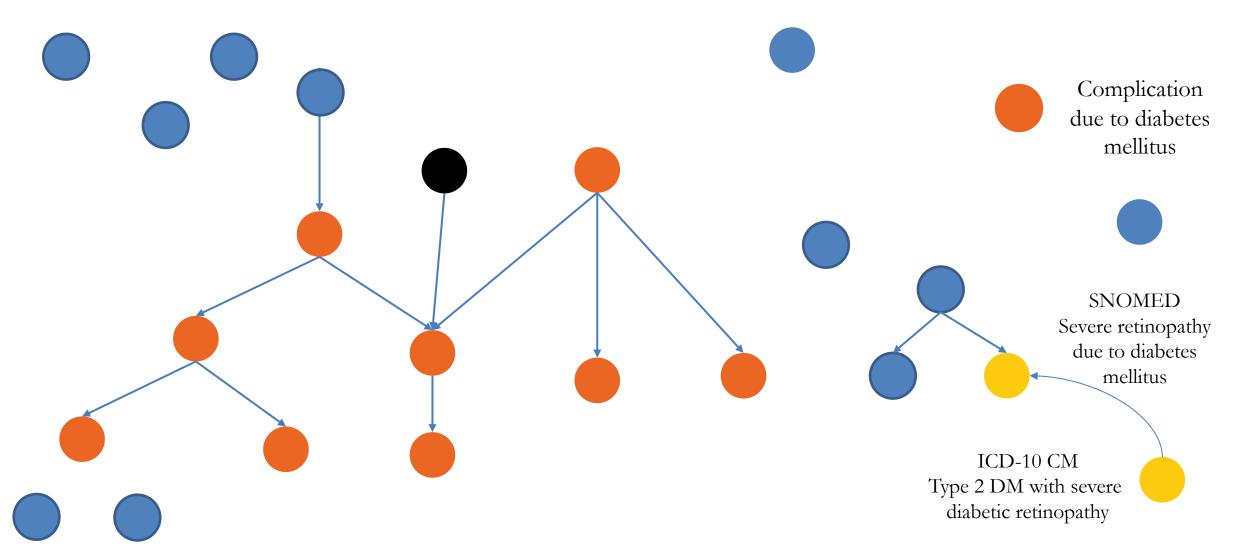


Check non selected descendants



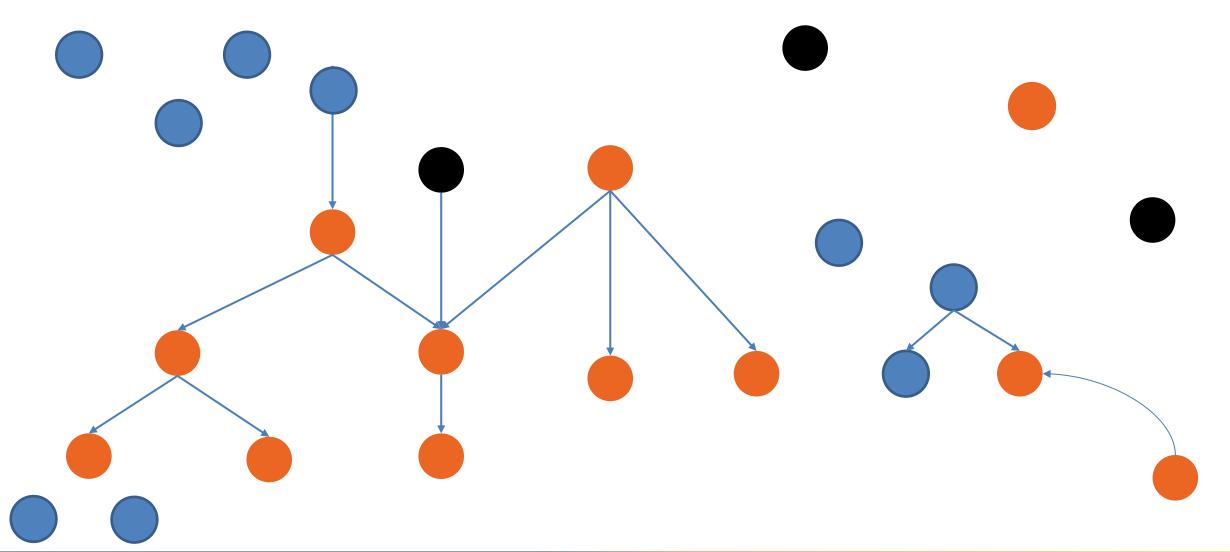


Check similar concepts from other places in hierarchy





Iterating until a comprehensive concept set





A principled algorithm for concept selection

- 1. Pick initial concept
- 2. Review and add descendants
- 3. Review and add parents
- 4. Review lexically similar concepts
- 5. Review and remove included concepts
- 6. Create concept set
- 7. Repeat steps 2 6 until the concept set is unchanged



PHenotype Observed Entity Baseline Endorsements (PHOEBE)

