



# APAC Community Call

August 21, 2025



# Agenda

- 2025 APAC Studies: Call for Data Partners
  - Regional Chapter Mid-year Updates: Australia, Malaysia
  - Deep Dive into OHDSI India's Developments
  - 2025 APAC Symposium Updates
-



# Call for Data Partners

2025 APAC Studies



# **Gastrointestinal Risk of GLP-1 Receptor Agonists versus SGLT-2 and DPP-4 Inhibitors in Type 2 Diabetes: A Multi-Database Observational Study**

Yongqi Zheng, Department of Epidemiology and Biostatistics, Peking University

Feng Sun, Department of Epidemiology and Biostatistics, Peking University



# Study Background

Pharmacologic  
Approaches to T2DM  
Treatment

Metformin: foundational glucose-lowering therapy, typically used as first-line agent  
GLP-1RAs: recommended as monotherapy or as add-on therapy to agents such as metformin in patients with T2DM and established ASCVD, CKD, or obesity, based on their cardiovascular and renal benefits

Evaluate  
Gastrointestinal(GI)  
Risks of GLP-1RAs  
Using Real-World  
Evidence:  
Rationale & Gap

**Safety Concern:** Delayed gastric emptying → possible gastroparesis, intestinal obstruction

**Conflicting Evidence:**

- Several large-scale observational studies (e.g., Sodhi et al., Nielsen et al.) have reported significantly increased risks of gastroparesis and intestinal obstruction
- Several studies (e.g., Gao et al., Ueda et al.) found no significant increase in gastrointestinal obstruction
- There have been no RCTs specifically designed to evaluate the gastrointestinal safety outcomes of GLP-1RAs

**Need for Real-World Evidence:**

- Prior studies limited by insufficient confounder adjustment
- Few multi-database studies on GI safety of GLP-1RAs



# Study Objectives

- Primary objective
  - Compare the risk of gastroparesis in T2DM patients initiating GLP-1RAs versus DPP-4 inhibitors or SGLT-2 inhibitors
  - Compare the risk of intestinal obstruction in T2DM patients initiating GLP-1RAs versus DPP-4 inhibitors or SGLT-2 inhibitors
- Secondary objective
  - Compare the risk of acute pancreatitis and nonalcoholic fatty liver disease (NAFLD) in T2DM patients initiating GLP-1RAs versus DPP-4 inhibitors or SGLT-2 inhibitors, as secondary outcomes to provide a broader assessment of digestive system safety



# Study Design

Analytic use case	Type	Structure
Population-level effect estimation	Comparative effectiveness	Does exposure to <b>GLP-1RAs</b> have a different risk of experiencing <b>gastroparesis or intestinal obstruction</b> within <b>end of continuous observation</b> , relative to <b>DPP4-i or SGLT2is</b> among the population with <b>type 2 diabetes and history of metformin</b> ?

## Population

### Inclusion

- Adults ( $\geq 18$  years) with T2DM
- $\geq 365$  days prior observation
- $\geq 90$  days prior metformin use

### Exclusion

- T1DM or secondary diabetes
- Prior exposure to study drugs or anti-diabetic exposure
- No prior insulin use or combo initiation
- History of pancreatitis, digestive system cancer, or abdominal surgery
- Renal dialysis, renal transplantation or end stage renal disease

**Target:** GLP-1RAs

**Comparator:** SGLT-2 inhibitors, DPP-4 inhibitors

### Outcome:

#### Primary outcome:

- Gastroparesis
- Intestinal obstruction

#### Secondary outcome:

- Acute Pancreatitis
- NAFLD



# CohortDiagnostics

- URL of CohortDiagnostics package:
  - <https://github.com/ohdsi-studies/2025APACStudy-Peking/tree/master/CohortDiagnostics>
- For troubleshooting:
  - Open an issue at <https://github.com/ohdsi-studies/2025APACStudy-Peking/issues>
- Send your results to:
  - Yongqi Zheng [zyq4664@pku.edu.cn](mailto:zyq4664@pku.edu.cn)
  - OHDSI APAC [apacsymposium@ohdsi.org](mailto:apacsymposium@ohdsi.org)
- Join us at our study Teams channel: [2025 APAC Study 2 - Peking](#)





# **Association Between Fasting Plasma Glucose Levels and Annual Hospitalization Days: A Multicenter Study Using the OHDSI Framework**

Ph.D. Jianying Gu, Zhongshan Hospital, Fudan University

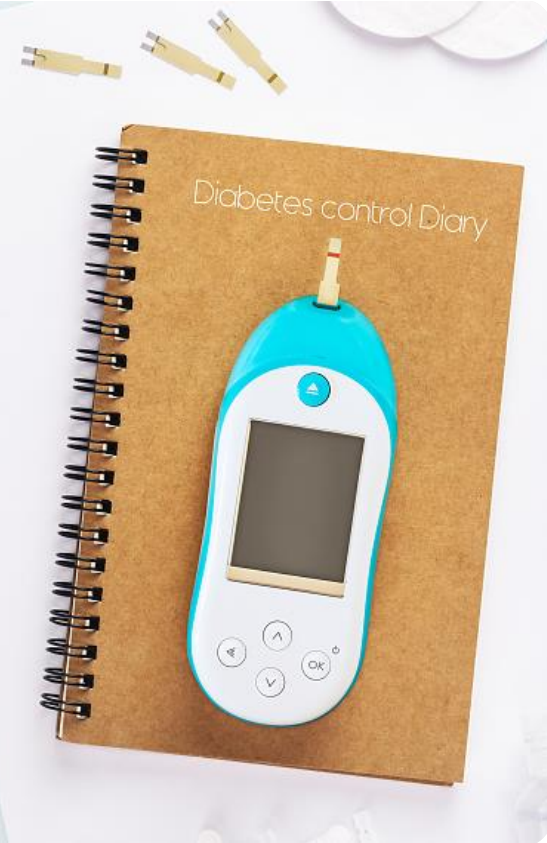
Ph.D. Lei Liu, Fudan University



# Study Background

## Fasting Plasma Glucose Levels

- Abnormal FPG levels relate to metabolic health, diabetes, and increased healthcare resource use.
- High levels can cause disorders, inflammation, and exacerbate chronic diseases, potentially leading to longer hospital stays



**OHDSI Framework about FPG:** Investigates association between FPG levels and annual hospitalization days using OHDSI framework, to inform glucose management and resource allocation strategies



# Study Objectives

## Problem statement— Exploring the Relationship FPG Levels and Hospitalization

✓ Evaluating how glucose abnormalities affect hospitalization days, aiming to quantify their impact on healthcare resources utilization

✓ Standard multicenter data in OHDSI to understand the relationship between plasma glucose levels and hospitalization days. This uses OHDSI's multicenter data to study how plasma glucose levels relate to the length of hospital stays, identifying trends for better care planning





# Study Design

The study population will be defined as:

1. Patients aged  $\geq 18$  years, of any gender.
2. A documented fasting plasma glucose (FPG) measurement during inpatient stay, including a valid numerical value and timestamp, expressed in mg/dL or mmol/L, and with FPG  $\neq 0$ .

**Patients must also have complete hospitalization records with both admission and discharge dates.**

**Additionally, patients must also meet one or both of the following criteria:**

- Admission to a specified clinical department (specialty).
- More than one hospitalization episode.

## **Exclusions:**

Incomplete hospitalization records or hospitalization dates falling outside the past 5 years.



# Preliminary Analysis to Determine Feasibility

## 1. FPG only

Analytic use case	Type	Structure
Clinical characterization	Disease Natural History	Amongst patients <b>with at least one fasting plasma glucose (FPG) measurement within a fixed continuous observation period starting from the event</b> , what are the patient's characteristics from their medical history?
Clinical characterization	Disease Natural History	Amongst patients <b>who experience an inpatient or ER visit accompanied by glucose measurement during the same clinical encounter</b> , what are the patient's characteristics from their medical history?

## 2. Glucose measurement in hospital



# CohortDiagnostics

- URL of CohortDiagnostics package:
  - <https://github.com/ohdsi-studies/2025APACStudy-Fudan/tree/master/CohortDiagnostics>
- For troubleshooting:
  - Open an issue at <https://github.com/ohdsi-studies/2025APACStudy-Fudan/issues>
- Send your results to:
  - Changran Wang [crwang@fudan.edu.cn](mailto:crwang@fudan.edu.cn), Jiaqi Liu [liu.jiaqi@zs-hospital.sh.cn](mailto:liu.jiaqi@zs-hospital.sh.cn)
  - OHDSI APAC [apacsymposium@ohdsi.org](mailto:apacsymposium@ohdsi.org)
- Join us at our study Teams channel: [2025 APAC Study 1 - Fudan](#)



# 2025 OHDSI APAC Symposium

December 6-7 • Shanghai Jiao Tong University, China







# Event Landing Page



Now Available!!





# Agenda (*Tentative*)

## Day 1 (Dec 6)

### Tutorial Sessions

- Introduction of OHDSI/OMOP
- OMOP CDM and Vocabulary
- ETL
- Analytics

## Day 2 (Dec 7)

### Updates & Studies

- OHDSI APAC and Regional Chapter updates
- 2025 APAC Studies: Overviews and Results
- Real-world Data Developments in China

### Discussions & Presentations

- Cross-community Panel Discussion
- Collaborator Showcase: Poster Presentations and Lightning Talks



# Collaborators Showcase

## 2025 OHDSI APAC Collaborator Showcase Brief Report Submission Form

Thank you for your interest in the 2025 OHDSI APAC Collaborator Showcase! We are delighted that you are considering joining our research community and presenting your work at this year's symposium. The 2025 OHDSI APAC Symposium will be held in person **December 6-7** at the Shanghai Jiao Tong University in Shanghai, China.

Please take a few minutes to fill out this submission form to help the OHDSI APAC Scientific Review Committee better understand your work. The deadline to submit your brief report is **September 7**. You will receive a confirmation email of your responses upon completion. If the committee has selected your work to be presented at this year's symposium, you will be notified via email by **October 17**.

Should you need to change your responses to any of the questions on this form, please click on the "Edit response" button in the confirmation email you received. Should you need to revise your brief report, please email [apacsymposium@ohdsi.org](mailto:apacsymposium@ohdsi.org). Your submission will be removed, and you will need to submit again with the revised PDF.



**Aug 4 – Sep 7**

**Sep 8 – Oct 6**  
**Oct 7-9 APAC WG workshop**

**Oct 17**

**Dec 6-7**

**Abstract Submission**

**Review by Scientific  
Committee**

**Notification of  
acceptance**

**Symposium**



## **Regional Chapter Mid-year Updates**

**Malaysia – Steven Young**



# Updates from Malaysia

1. **MyGenom** has been introduced to the OMOP CDM framework and will begin incorporating the learnings into their genomic and health data initiatives, aligning with global data interoperability standards. Intention is to expand its capabilities into the area of pharmacogenomic.
  - Will be reaching out to the community / team with these experience.

## What is the **MyGenom** Project?

The MyGenom Project is an effort to understand the genetic makeup of Malaysians. It's our country's first large-scale study of population genomics, which means sequencing the DNA of Malaysian citizens from different races and ethnicities. In Phase I, 2,400 genomes will be sequenced from healthy Malaysian individuals, with the goal of reaching 10,000 genomes in Phase II. Over time, this project will create a Malaysia reference genomic dataset that reflects the diversity of our population and will be the key to bringing Precision Medicine to Malaysia.



# Updates from Malaysia (continue)

2. **National Institutes of Health (NIH)** is planning to establish its **Trusted Research Environment (TRE)** in **Q1/Q2 2026**. This TRE will model after OMOP CDM principles, emulating the success of TRUST framework, and NIH is currently applying for funding to support the project.
  - Will be reaching out to the community / team with these experience.
  
3. **Ministry of Health (MOH)** will be embarking on its **first OMOP CDM project** in **Q1 2026**, paving the way for broader adoption of standardized healthcare data models across the public healthcare sector. The first phase will start off with data from public clinics and followed by hospitals. These initiatives is being done in parallel to their nationwide effort in digitalizing clinics across the country.



# **Deep Dive into OHDSI India's Developments**

**Swetha J.**



# OHDSI India Chapter

*From Raw Data to Reliable Evidence*

# Agenda

**Breast Cancer Study Overview**

**FHIR Work Group Initiative**

**The CVD Patient Registry**

**Digital Health Centre of Excellence**



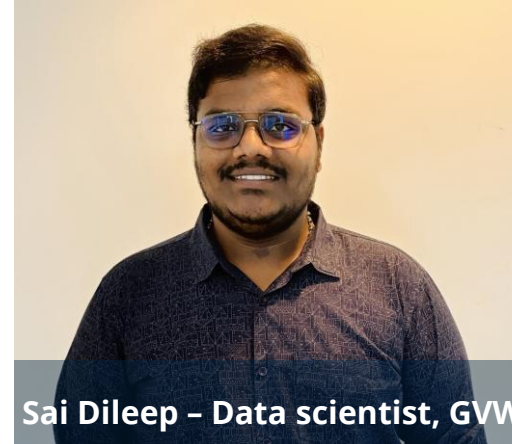
# Our Team



**Dr. Swetha Kiranmayi Jakkuva,**  
Service Leader Research enabling  
services, GVW



**Khansa Fathima – Assistant  
Professor at JSSAHER**



**Sai Dileep – Data scientist, GVW**



**Shreema – Data Engineer, GVW**



**Sanjay – Healthcare Data  
analyst, GVW**



**Sai Pattabhiram – Software  
Engineer, GVW**



**Vipina Keloth - Associate Research  
Scientist at Yale Biomedical  
Informatics and Data Science**

# Breast Cancer Study

## Problem Statement:

PDF discharge summaries from 2014 to 2022 at JSS Mysore University containing unstructured breast cancer data. This unstructured format presents challenges in accessing, standardizing, and analyzing data for research and clinical insights.

## Objectives:

Extract and structure clinically relevant data from the PDF discharge summaries using NLP techniques.

Standardize the extracted data to the OMOP Common Data Model (CDM) for interoperability and analysis.

Enable efficient data utilization for research and real-world evidence (RWE) generation in breast cancer care.

# Know our Data

**DISCHARGE SUMMARY**  
**DEPARTMENT OF PAIN AND PALLIATIVE CARE**

REG/IP NO : 1000002 / 80002	NAME : Patient 2
SEX : FEMALE	ADMITTED ON : 19-09-2017
BED NO : 5-F-52	DISCHARGE ON : 25-09-2017

**Description**  
K/C/O ca ovary post exploratory laparotomy status with post chemotherapy status.

**Diagnosis**  
Operated case of Ca Ovary -Metastatic adenocarcinoma?on palliative care

**History**  
c/o excess tiredness since 3 days  
c/o difficulty in passing with blood tinged stool.c/o oral ulcer and painful swallowing  
c/o vomiting  
h/o weight lossappetite - poor  
no cardiopulmonary symptomsno other comorbidities

**present medications-**  
tab morphine cr 10mg 2-0-2  
tab amixide-h 0-0-1  
tab prelogic 75mg 0-0-1

**Investigation Done**  
urea - 31, creatinine - 0.7, Na/K/Cl- 135/4.3/98 , hb-9.4,pcv- 29.3, tlc- 8820, platelet -3.59, blood group - BPOSITIVE, URINE ROUTINE-normal.

**Course in the Hospital :**  
patient was transfused with 1 pint PRBC, ENT examination done. was treated symptomatically with antiemetics, mucaine gel, MVI and continued with past medications.

<http://jschhis/backbone/ward/frmSumPreview.aspx?Action=Print&preview=yes&opt=1&Si...> 6/7/2023

frmSumPreview Page 2 of 2

REG/IP NO : 1000002 / 80002	NAME : Patient 2
-----------------------------	------------------

**Advice on Discharge-**

continue  
tab morphine cr 10mg 2-0-2 for 30days  
tab amixide-h0-0-1 for 15days  
tab prelogic 75mg 0-0-1 for 30days  
tab cremalax 0-0-1 7 days and sos  
tab pan - d 1-0-0 for 7 days and SOS  
thereafterensure powder 2tsp 1-1-1 a to z 2tsp 1-1-1

review in pain opd SOS.

<http://jschhis/backbone/ward/frmSumPreview.aspx?Action=Print&preview=yes&opt=1&Si...> 6/7/2023

Demographic

Medical history

Labs / Diagnostics

Procedures

Drugs

# Process Overview

**PDF Deidentification to remove PHI**

**Conversion of PDF to text**

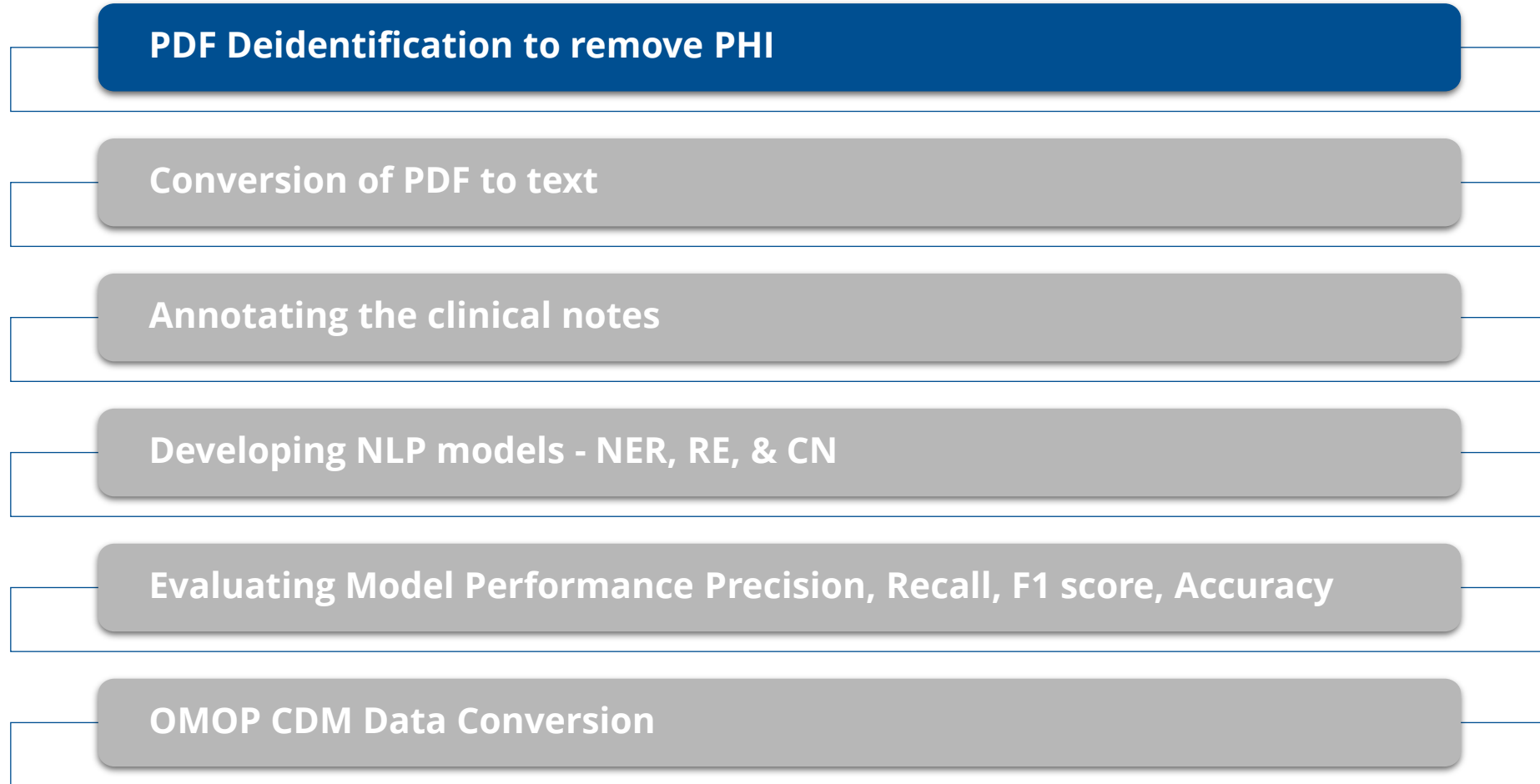
**Annotating the clinical notes**

**Developing NLP models - NER, RE, & CN**

**Evaluating Model Performance Precision, Recall, F1 score, Accuracy**

**OMOP CDM Data Conversion**

# Process Overview



# Deidentification

Id : 2fde18bf55ee24c03e7024a5b0b2721055739d02196dc7b7480ed5214ecaa8c5

## DISCHARGE SUMMARY

### DEPARTMENT OF MEDICINE

REG/IP NO : [REDACTED]	NAME : [REDACTED]
AGE : 30 Years	ADMITTED ON : 23-11-2014
SEX : FEMALE	DISCHARGED ON : 27-11-2014
BED NO : GM-F-01	UNIT HEAD : [REDACTED] M MBBS,MD,FACP (USA),FRCP
ADDRESS : [REDACTED]	
MYSURU	

#### GENERAL MEDICINE

**FINAL DIAGNOSIS:** Ca Breast with ?Secondaries to the Brain.

**PRESENTING SYMPTOMS:** Headache- 10 days

Weakness of left side of body-8 days back lasted for 10 min.

Fever- 2days

Diplopia -2 days

**PAST HISTORY:** K/C/O of Ca Breast 1 year back for which she got operated in month of Aug 2013 followed by 2 cycles of chemotherapy.

#### EXAMINATION:

GPE- PR : 90bpm BP : 118/70 mmhg

CVS- NAD

RS - NAD

PA- NAD

CNS-Conscious and oriented.HMF-normal

Power Rt UL and LL-5/5

Lt UL and LL-4/5

Plantar Rt-flexor lt-extensor

**INVESTIGATIONS:** Enclosed

**TREATMENT GIVEN:**Inj Mannitol 100ml i.v 1-1-1

Tab Strocit-P 1-0-1

Tab Ultracet 1-0-1

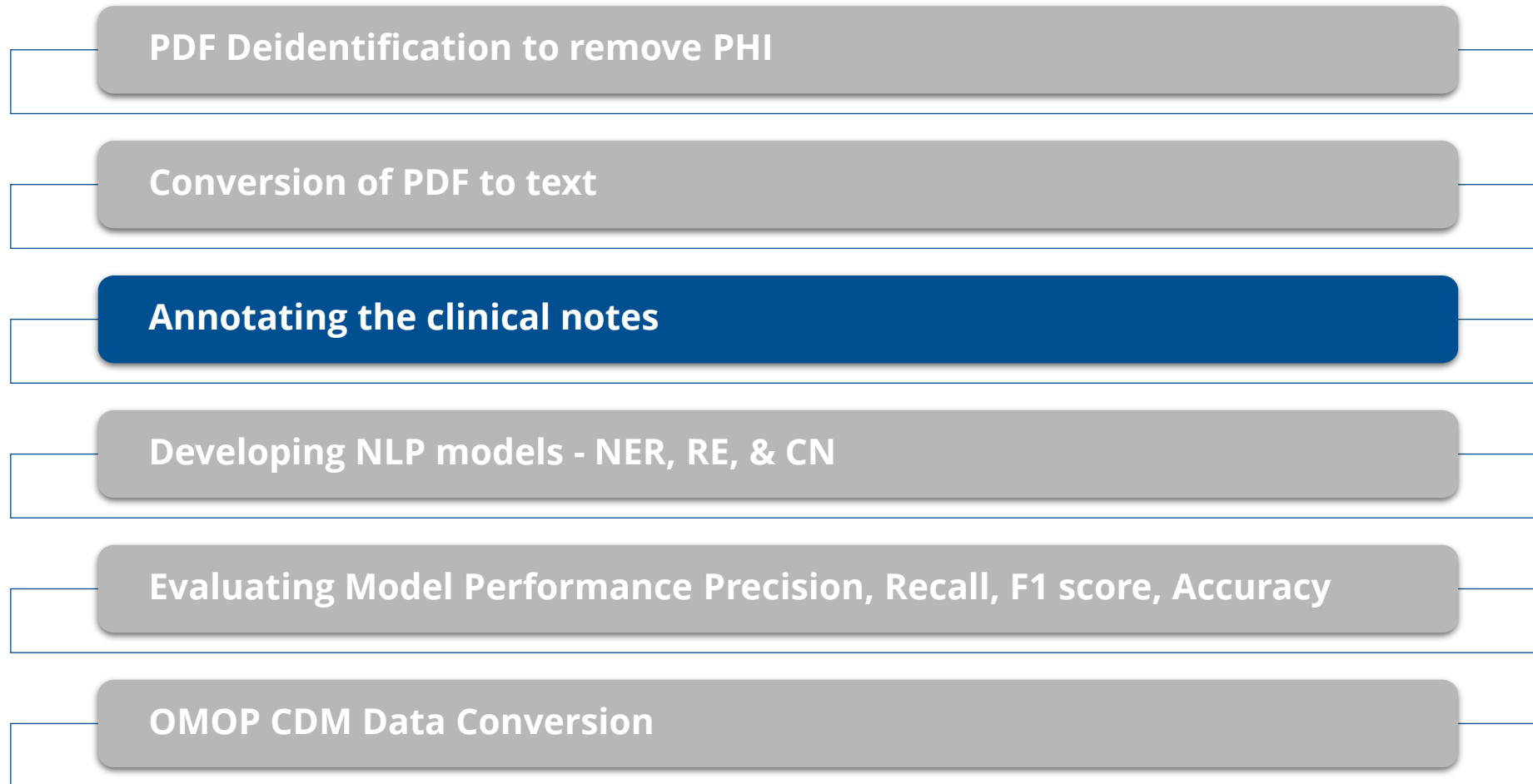
Tab Rantac 150 mg 1-0-1

Tab Eptoin 100mg 1-0-2

Inj Dexona 8mg stat and 4mg 1-1-1

A specialized deidentification tool is used to mask the PHI from PDF discharge summaries, ensuring patient privacy and compliance with data security standards.

# Process Overview



# Develop Annotation Guidelines

Defined the main entities and modifiers based on the purpose of annotation.

Entity Name	Definition
Problem	The abnormal condition that happens physically or mentally to a patient.
Procedure	A medical process or method used to perform a test or treatment of a problem.
Drug	Generic or brand name of a single medication or a collective name of a group of medication.
Test	A medical procedure performed (i) to detect or diagnose a problem, (ii) to monitor diseases, disease processes, and susceptibility, or (iii) to determine a course of treatment.

Modifier Name	Definition
Severity	The degree of intensity of a clinical condition
Conditional	A phrase that indicates the problems existing in a certain situation
Uncertain	A measure of doubt related to a problem
<u>Bodyloc</u>	The location on the body where the observation is present
Course	The development or alteration of a problem
Temporal	Calendar date, time, or duration related to a problem
Negation	The phrase that indicates the absence of an entity
Problem Status	Indicates the temporal status of the problem, distinguishing between current and past conditions.



# Develop Annotation Guidelines

## Basic annotation rules

Example - Annotate demographic data such as gender (male/female) and age information

## Specific guidelines for each main entity

**Problem - Definition:** The phrases containing the observations made by the patient or clinician about the patient's body or mind as abnormal or caused by a disease are considered as a "Problem".

## Special cases and document as you annotate

Chemotherapy should be annotated as a procedure entity, while the specific drugs used in the chemotherapy should be annotated as drug entities.

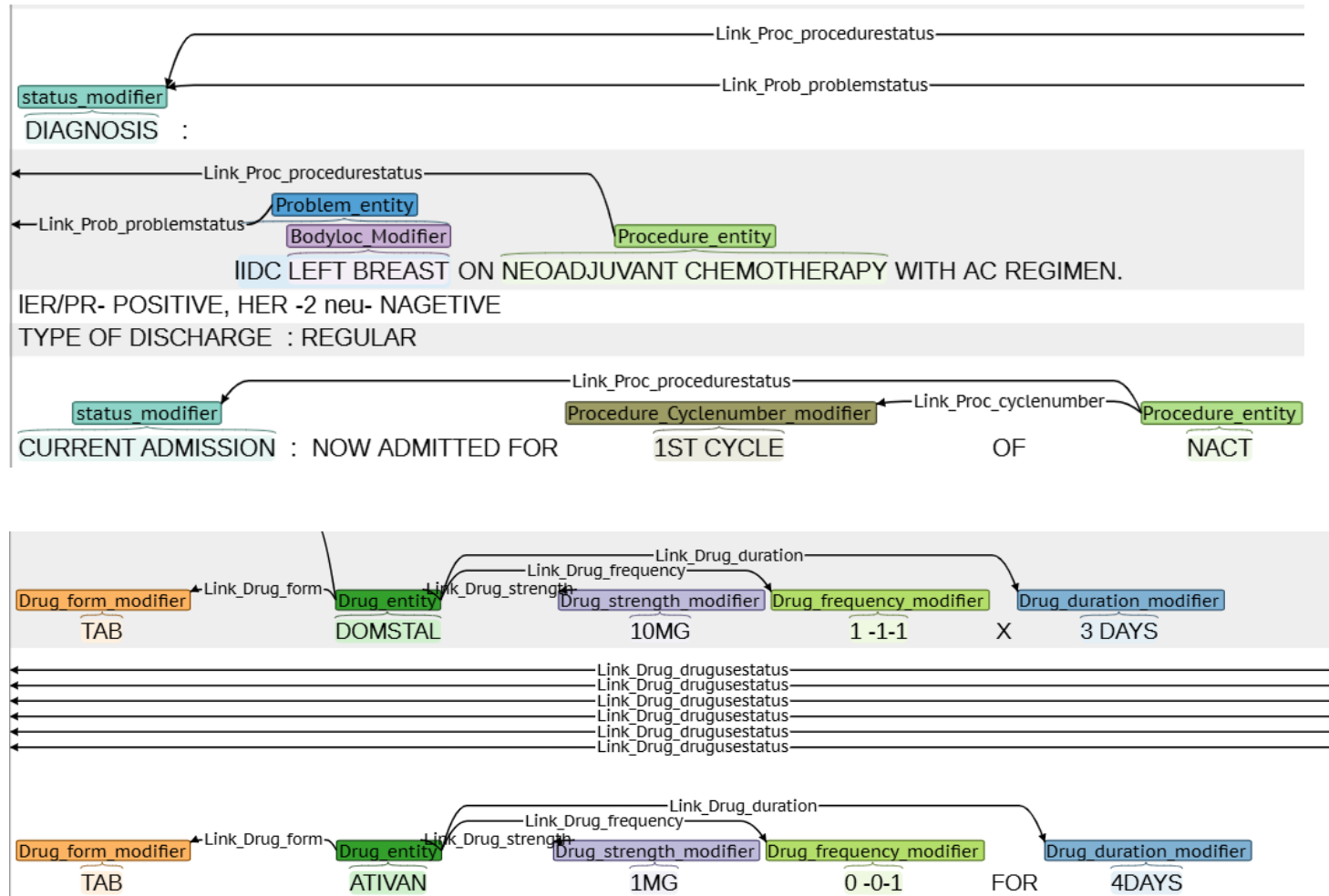
## The modifiers

When an ENTITY is described by a MODIFIER, the MODIFIER provides additional information about the ENTITY

## Span relationship between modifiers and entities

After annotating the entity, attribute the relationships between the associated modifiers.

# Annotations



## Annotation Schema

You need to prepare the annotation schema.

### Schema: Breastcancer\_Annotation

- 27 Entity Tags

Demographic\_entity Problem\_entity  
Procedure\_entity Drug\_entity Test\_entity  
Problem\_Severity\_Modifier  
Problem\_Conditional\_Modifier  
Problem\_Uncertain\_Modifier Bodyloc\_Modifier  
Problem\_Course\_modifier Temporal\_modifier  
Negation\_modifier status\_modifier  
Procedure\_Cyclenumber\_modifier  
Drug\_strength\_modifier Drug\_dosage\_modifier  
Drug\_duration\_modifier Drug\_form\_modifier  
Drug\_frequency\_modifier Drug\_route\_modifier

Test\_Labvalue\_modifier

Test\_Referencerange\_modifier

Demographic\_age\_modifier

Demographic\_sex\_modifier

Demographic\_uniqueid\_modifier

Demographic\_Admissiondate

Demographic\_Dischargedate

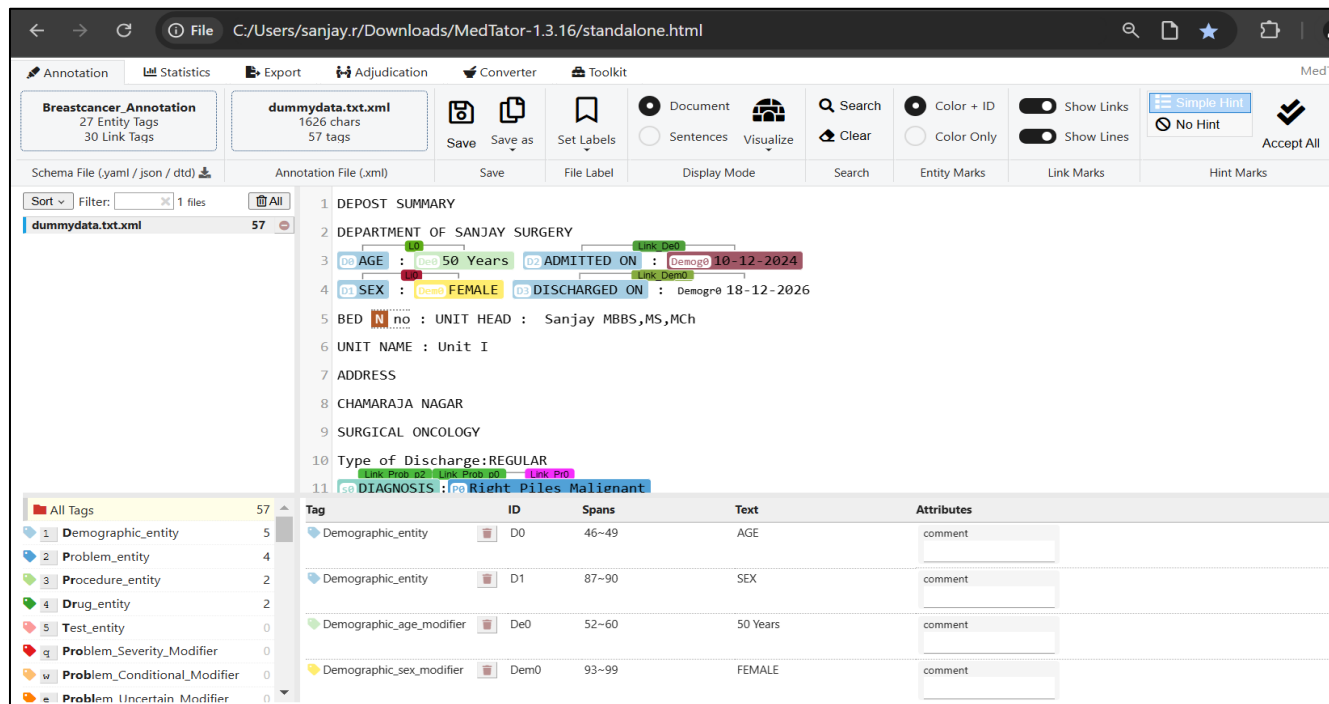
- 30 Link Tags Link\_Demo\_age Link\_Demo\_sex

Link\_Demo\_uniqueID Link\_Prob\_severity

# Annotation Tool & Methodology



GitHub [/OHNLPMedTator](https://github.com/OHNLPMedTator)



The screenshot displays the MedTator web application interface. The top navigation bar includes tabs for Annotation, Statistics, Export, Adjudication, Converter, and Toolkit. The main content area shows a medical text document with various entities highlighted and labeled. A left sidebar lists all tags and their counts. A bottom table provides details for specific tags.

Tag	ID	Spans	Text	Attributes
Demographic_entity	D0	46~49	AGE	comment
Demographic_entity	D1	87~90	SEX	comment
Demographic_age_modifier	De0	52~60	50 Years	comment
Demographic_sex_modifier	Dem0	93~99	FEMALE	comment

## Results:

Inter-Annotator Agreement (IAA) was measured across different batches of annotated files. The results are as follows:

- IAA (1):** Annotator A vs. Annotator B (Batch of 25 files) - **94.71%**
- IAA (2):** Annotator A vs. Annotator C (Batch of 75 files) - **96.12%**
- IAA (3):** Annotator A vs. Annotator D (Batch of 75 files) - **93.10%**
- IAA (4):** Annotator A vs. Annotator D (Batch of 125 files) - **94.32%**

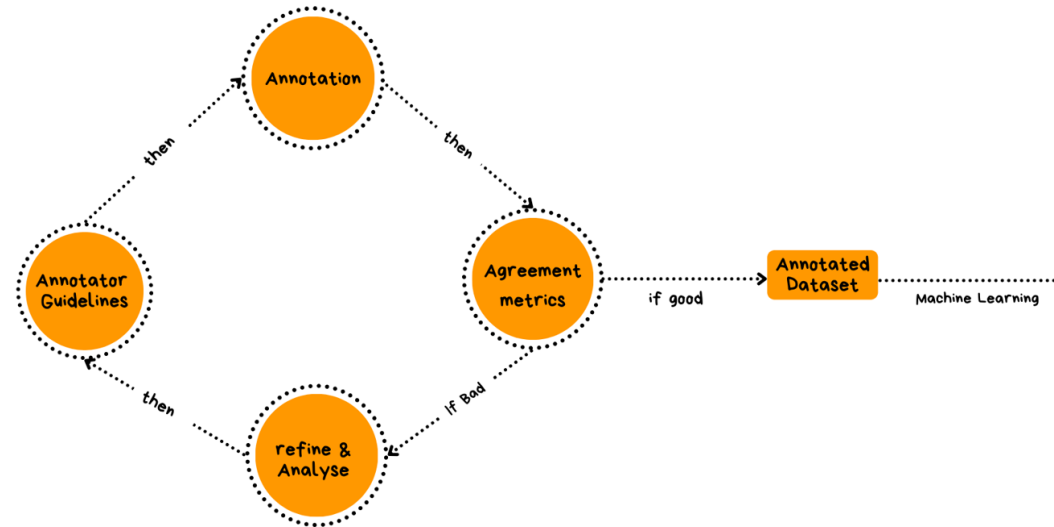
## Error Analysis :

Error analysis was conducted by Annotator A to maintain quality control and ensure accurate annotations.

# Inter-annotator Agreement (Adjudication)



- Cohen's Kappa
- Krippendorff's alpha



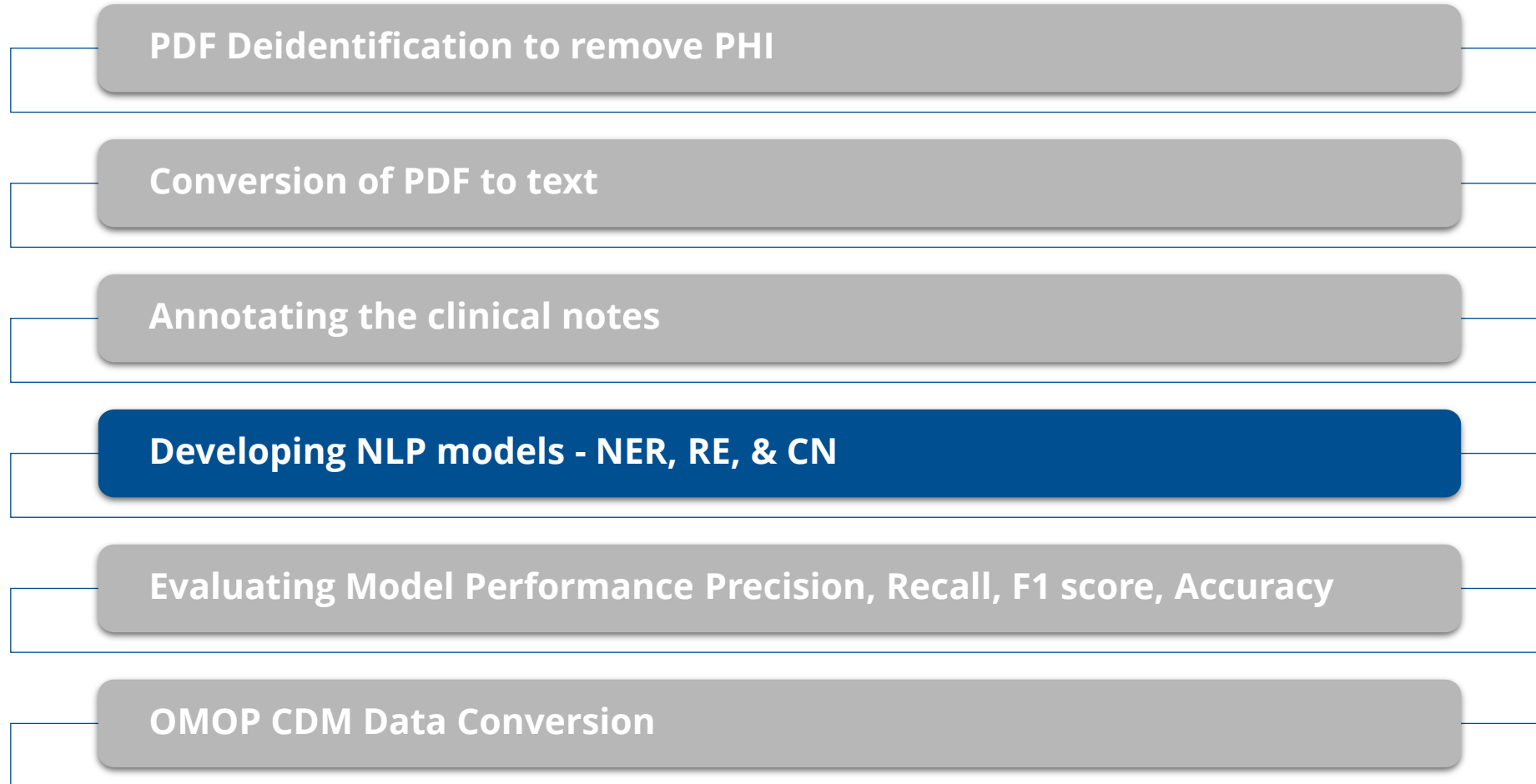
$$\text{Precision} = \frac{\text{True Positive}}{\text{True Positive} + \text{False Positive}}$$

$$\text{Recall} = \frac{\text{True Positive}}{\text{True Positive} + \text{False Negative}}$$

$$\text{F1} = 2 \times \frac{\text{Precision} \times \text{Recall}}{\text{Precision} + \text{Recall}}$$

1	tag_name	F1	precision	recall	TP	FP	FN	
2	Overall	0.8866	0.8337	0.9468	2436	486	137	
3	Demographic_entity	0.9540	0.9120	1.0000	114	11	0	
4	Problem_entity	0.6405	0.5024	0.8833	106	105	14	
5	Procedure_entity	0.7547	0.6557	0.8889	80	42	10	
6	Drug_entity	0.8931	0.8224	0.9771	213	46	5	
7	Test_entity	0.9507	0.9450	0.9564	395	23	18	
8	Problem Severity	1.0000	1.0000	1.0000	1	0	0	

# Process Overview



# Data extraction from the discharge summaries



## Named entity recognition

*Identifies 27 modifiers across demographics, problems, procedures, tests, and drugs, including sub-modifiers.*



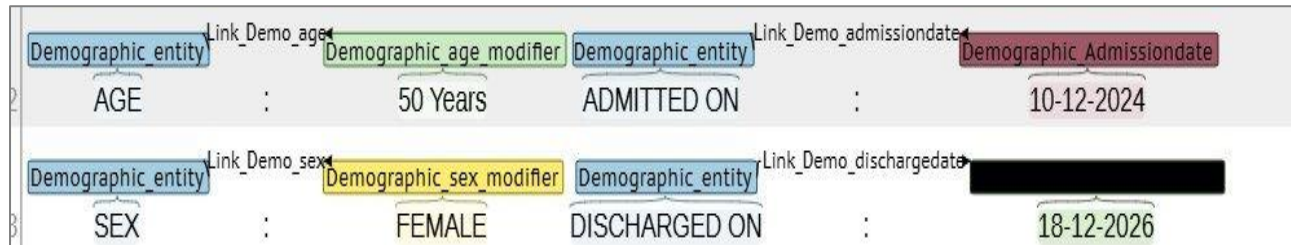
## Relation extraction

*Maps relationships like historical vs. current problems and determines severity.*

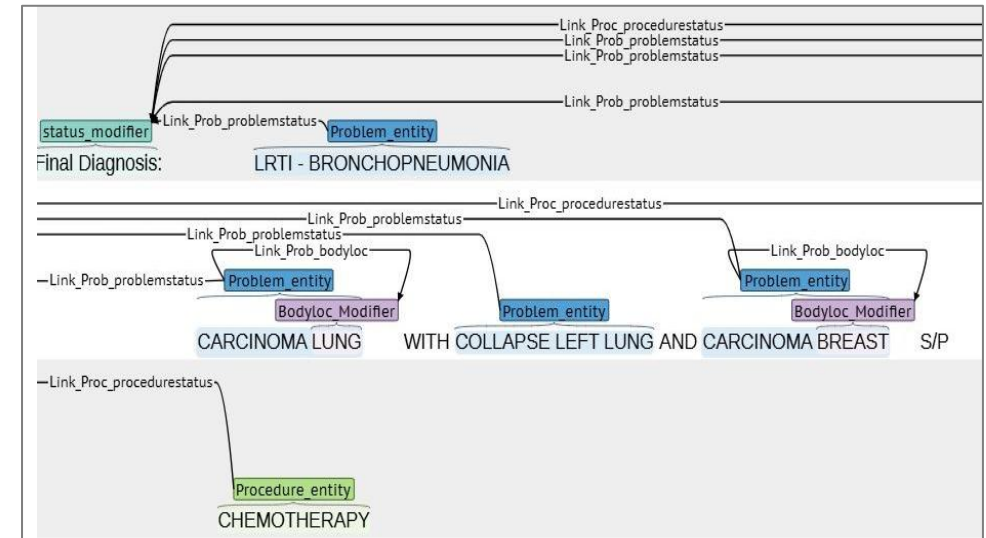


## Concept Normalization and Vocabulary Mapping

*Normalizes and aligns entities to OMOP Concept IDs for standardization*



TRAIN:VALIDATE:TEST – 70:10:20



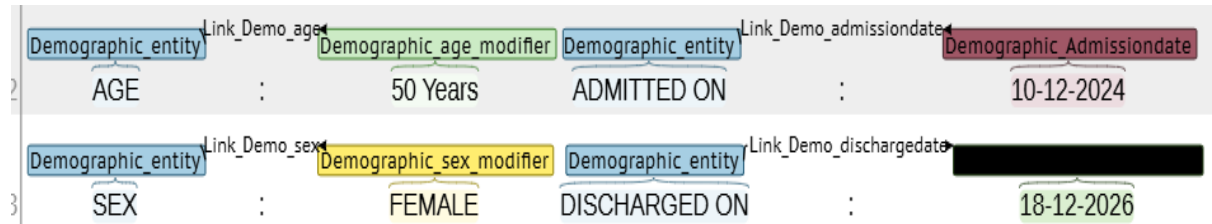
# Named Entity Recognition

## BERT Model

Identifying the boundaries of entities

Identifying the semantic categories of entities

## BIO ( B – Beginning I – Inside O – Outside ) Tagging






### Token

### Label

AGE	B-Demographic_entity
:	O
50	B-Demographic_age_modifier
Years	I-Demographic_age_modifier
ADMITTED	B-Demographic_entity
ON	I-Demographic_entity
:	O
10-12-2024	B-Demographic_Admissiondate
SEX	B-Demographic_entity
:	O
FEMALE	B-Demographic_sex_modifier
DISCHARGED	B-Demographic_entity
ON	I-Demographic_entity
:	O
18-12-2026	B-Demographic_Dischargedate

# Evaluation

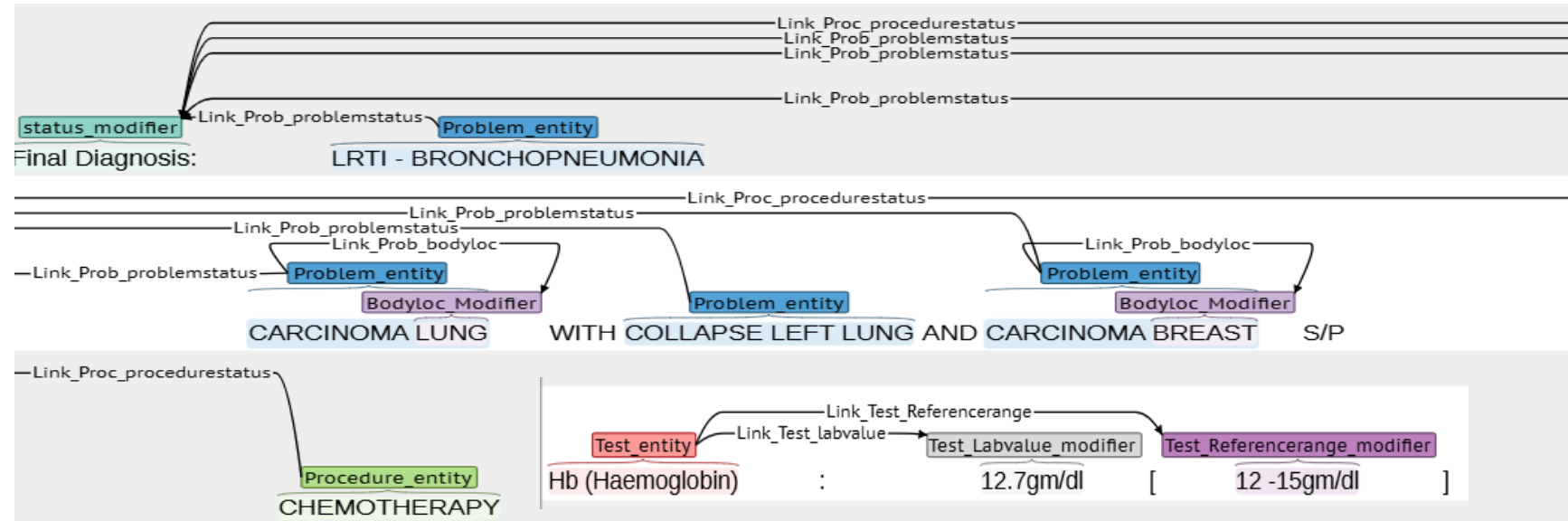
Text	Gold Annotation	Annotation	Result
The patient was diagnosed with <b>breast cancer</b>	breast → B-Problem cancer → I-Problem	breast → B-Problem cancer → I-Problem	 <b>Exact Match</b> (same entity type + same boundaries)
The patient was diagnosed with <b>breast cancer</b>	breast → B-Problem cancer → I-Problem	cancer → B-Diagnosis	 <b>Relaxed Match</b> (entity type matches, boundaries overlap but not identical)
The patient was diagnosed with <b>breast cancer</b>	breast → B-Problem cancer → I-Problem	breast → B-BodyPart	 <b>Not a Match</b> (entity type mismatch, even though text overlaps)



# Relationship Extraction

## BERT Model

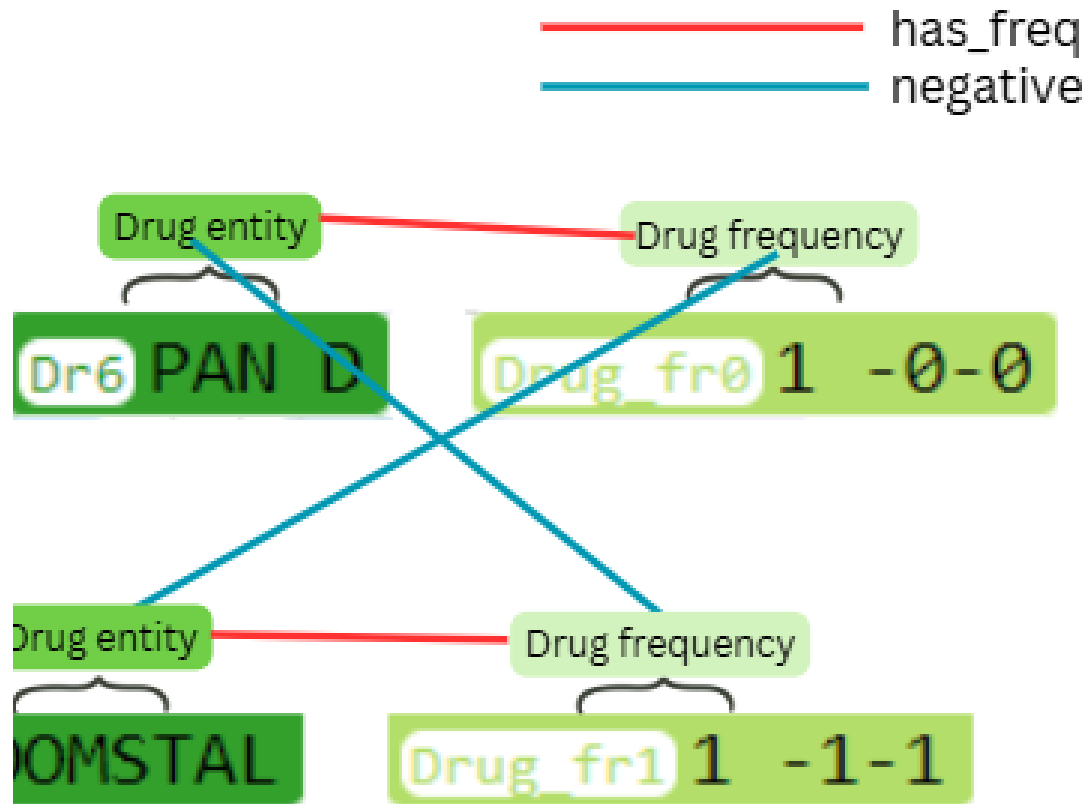
- Identify the semantic relations between entities.
- Determine the direction and context of the relation



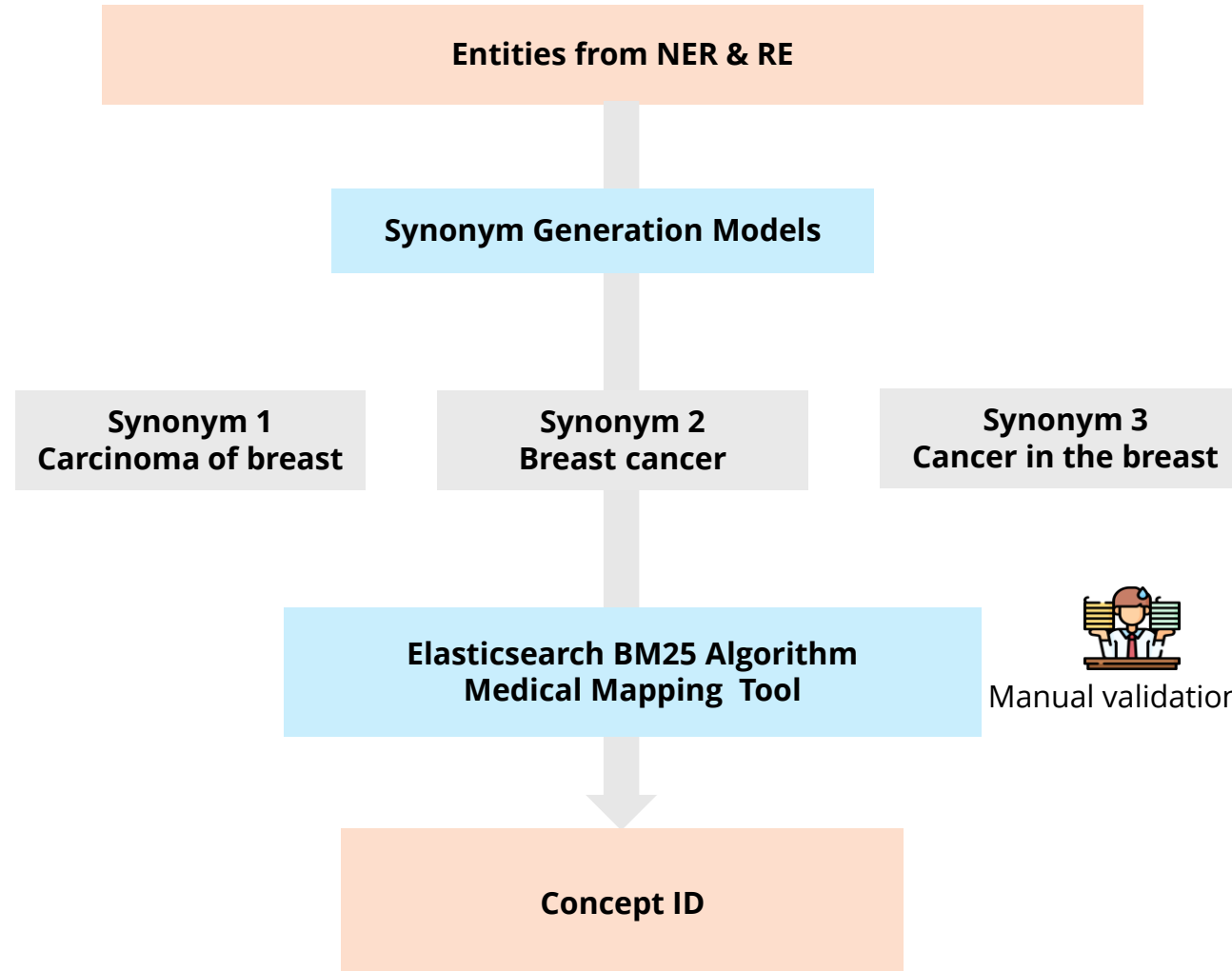
Entity	Attribute Value	Relationship
Carcinoma	Breast	Has_attribute
LRTI -Bronchopneumonia	Final Diagnosis	Has_attribute
Hb(Haemoglobin)	12.7gm/dl	Has_attribute

# Evaluation

- Sentence classification of 2 classes:
- [Has\_attribute, Negative]



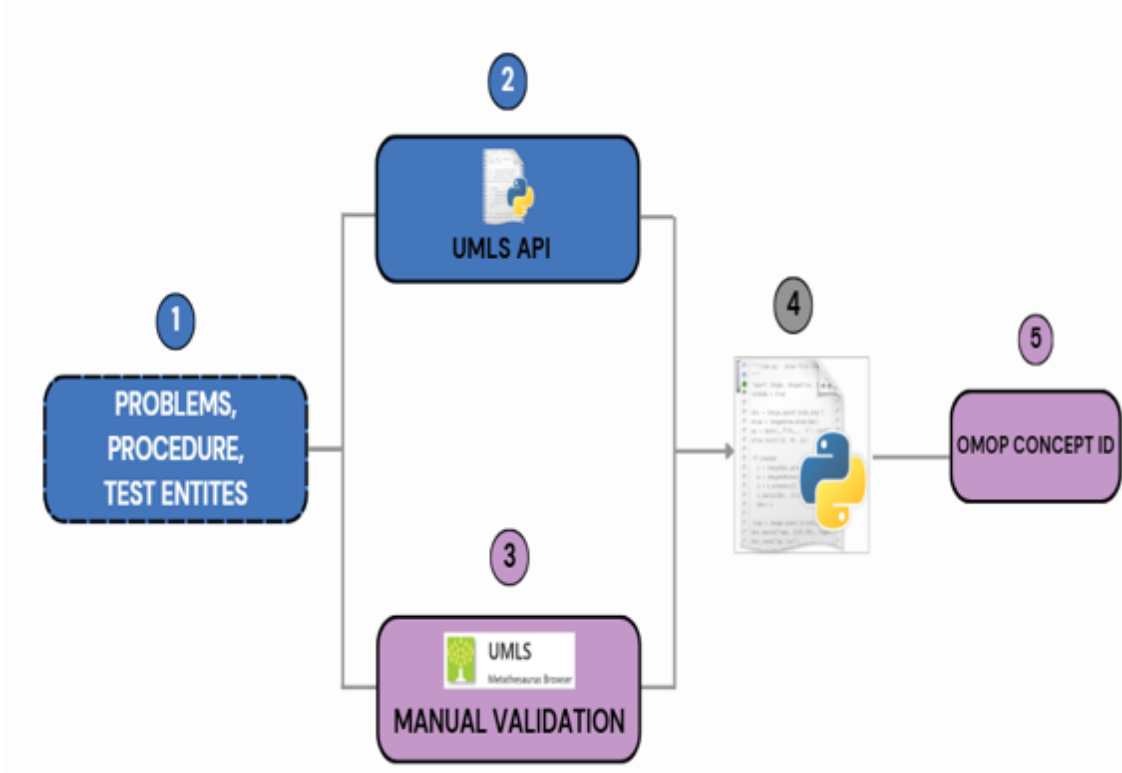
# NLP Step 3 - Concept Normalization & Vocabulary Mapping



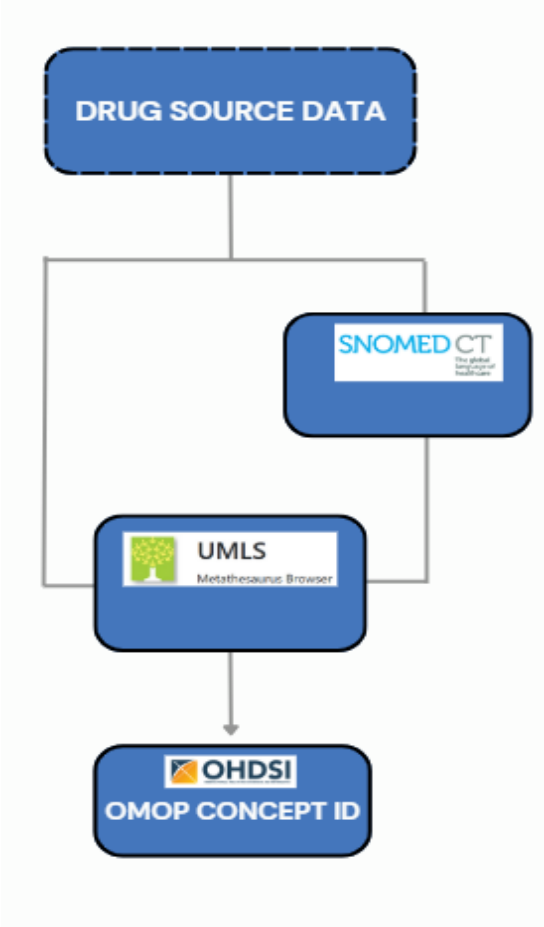
Manual validation

# NLP Step 3 - Concept Normalization & Vocabulary Mapping

## Clinical Text



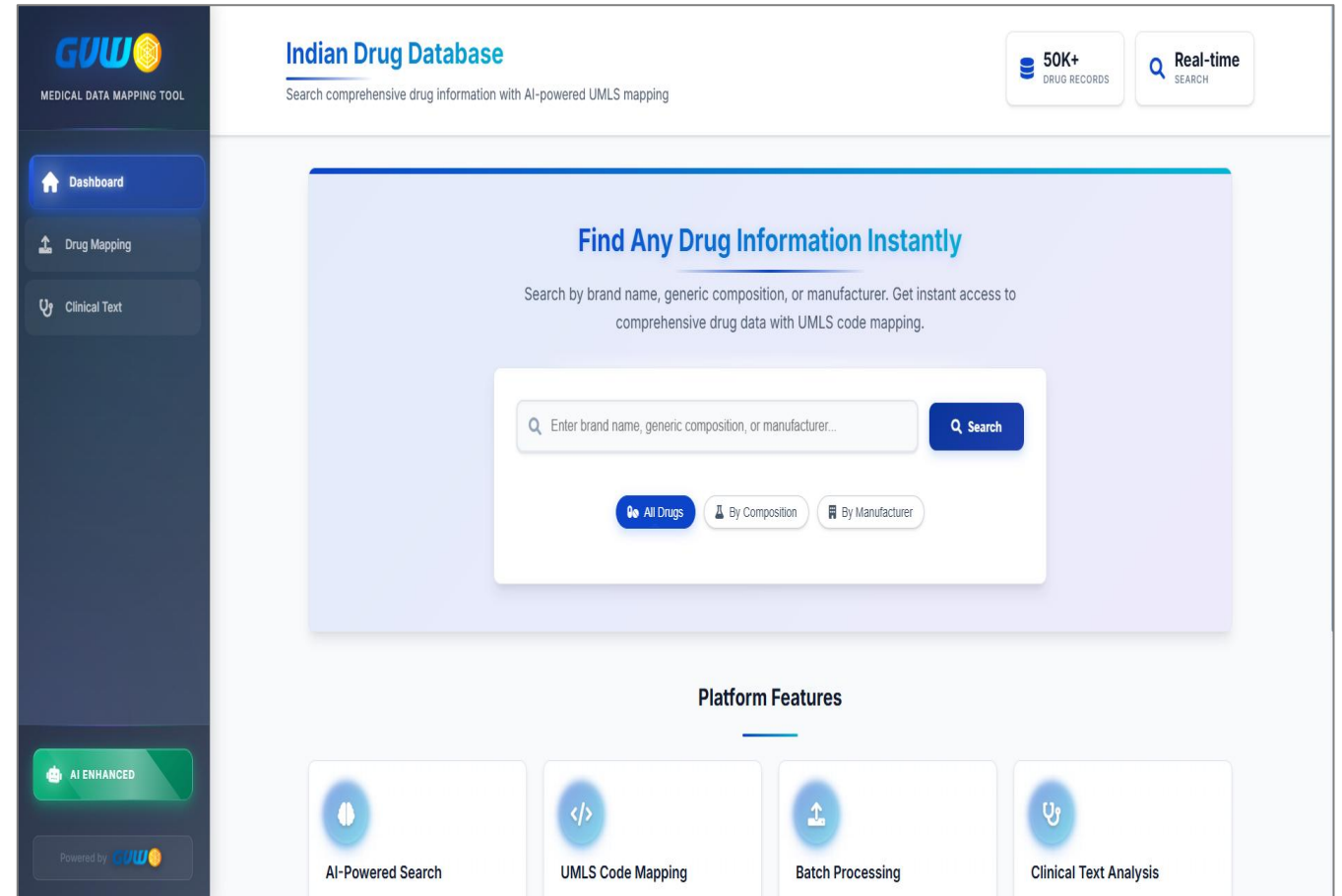
## Drugs



# Medical Data Mapping Tool

## Features –


- Maps terms to UMLS, RxNorm, and SNOMED CT.
- High accuracy with Gemini AI.
- Manual review and correction option.
- Auto spell-check and standardization.
- Identifies and standardizes abbreviations
- Maps Indian brand names to generics.
- Scalable for large datasets.



# Medical Data Mapping Tool - Input

## Input

drug	strength
Pantoprazole	40 mg
Dolo	650 mg
Atorvastatin	40 mg
Ultracet	
Aceclofenac	
Telma	40 mg
Terbinafine	
Cetirizine	10mg
Atorva	20 mg
Metformin	500 mg
Pan	40 mg
Paracetamol	650 mg



MEDICAL DATA MAPPING TOOL

- Dashboard
- Drug Mapping**
- Clinical Text

AI ENHANCED

### Drug UMLS Mapping

Upload pharmaceutical data for AI-enhanced UMLS code extraction

5x

FASTER PROCESSING

95%

ACCURACY RATE

Drop your file here or click to browse

Supports CSV, XLSX, XLS files up to 16MB

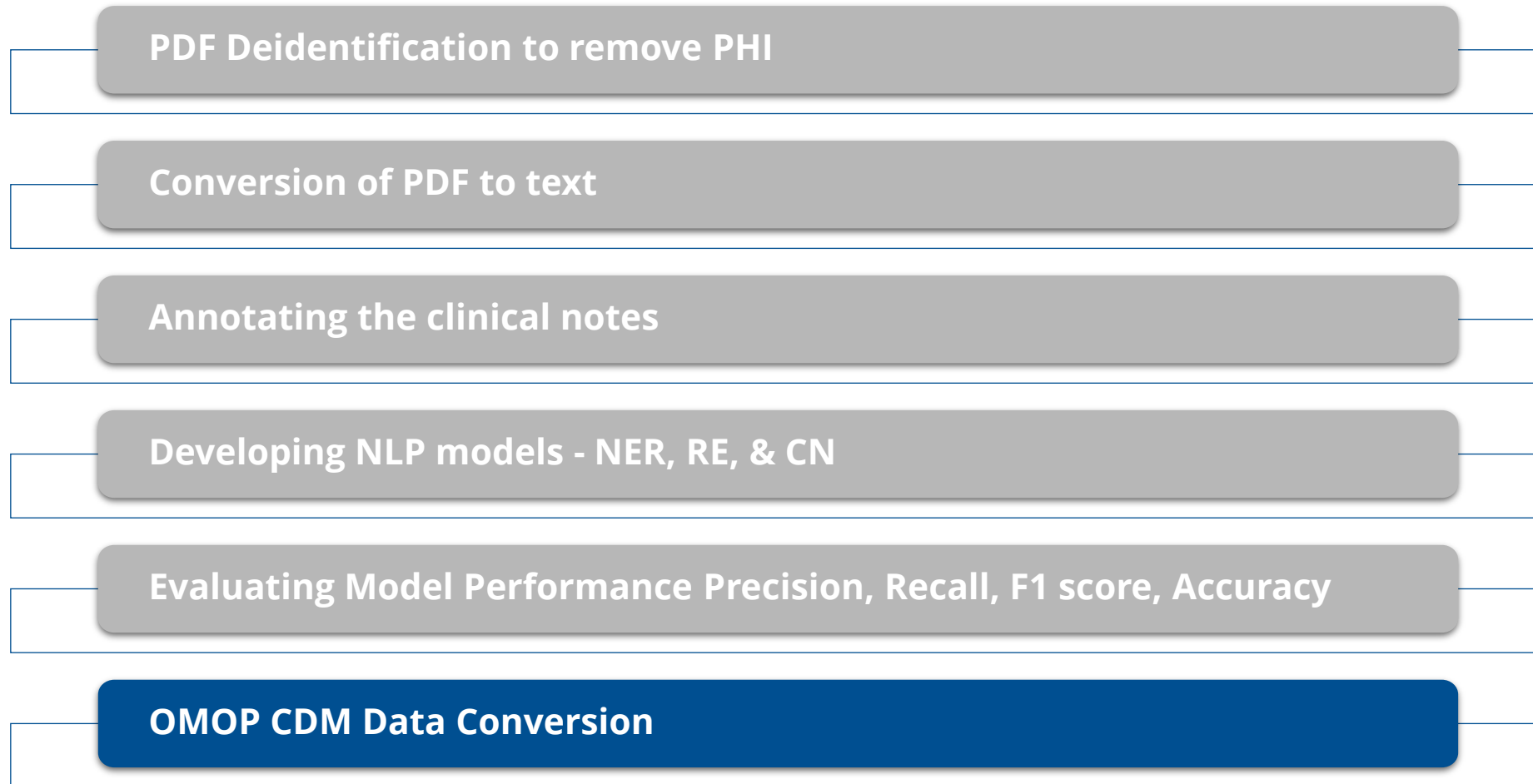
sample\_drugs1.csv

Preview & Configure

# Medical Data Mapping Tool - Output

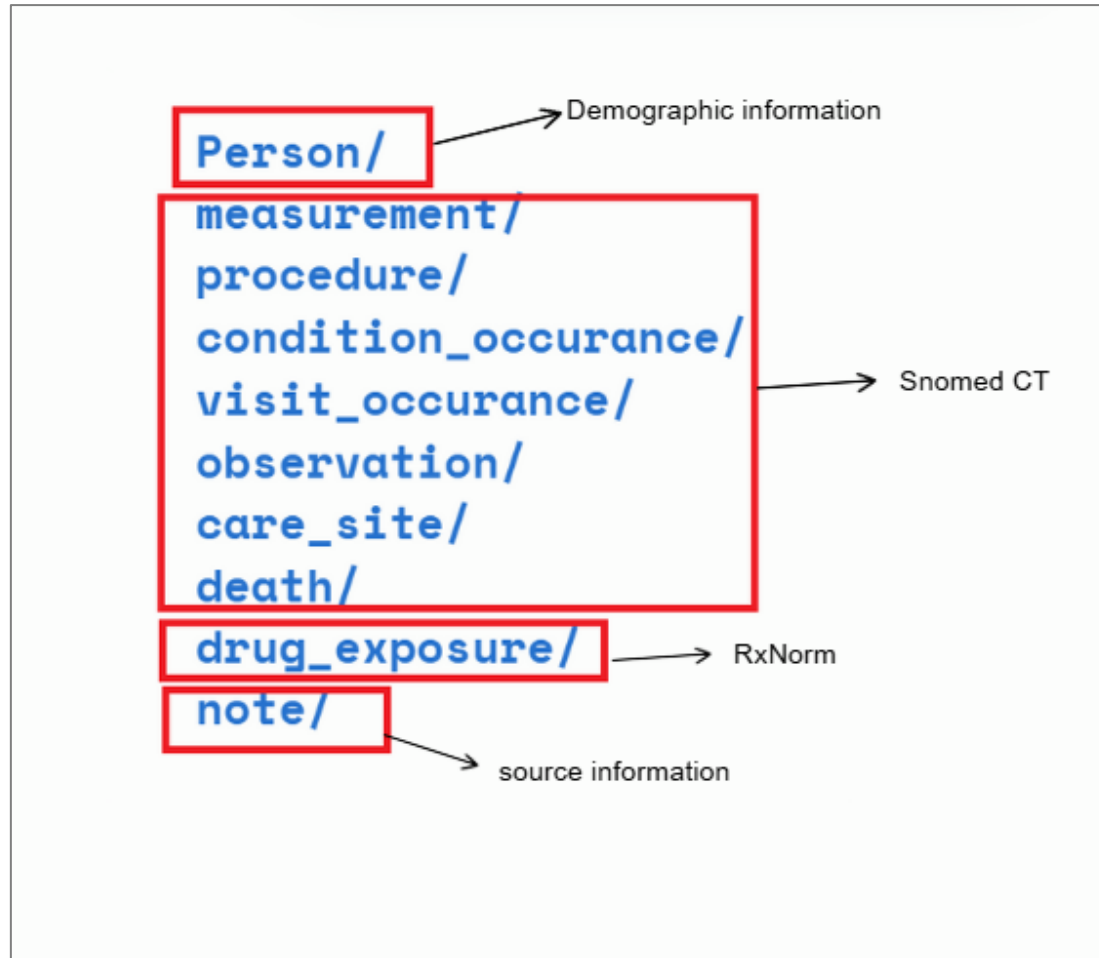
Term	CUI	Name	Source	Source Code	Semantic Types
pantoprazole 40mg	C1126048	pantoprazole 40 MG	RXNORM	330396	Clinical Drug
Dolo 650mg	C0691095	acetaminophen 650 MG Oral Tablet	RXNORM	198444	Clinical Drug
Atorvastatin 40mg	C1667395	atorvastatin 40 MG	RXNORM	597983	Clinical Drug
Ultracet	C1253063	acetaminophen / tramadol Oral Tablet	RXNORM	378712	Clinical Drug
aceclofenac	C0050403	aceclofenac	RXNORM	16689	Organic Chemical Pharmacologic Subs
telma 40mg	C0990502	telmisartan 40 MG	RXNORM	316764	Clinical Drug
terbinafine	C0076110	terbinafine	RXNORM	37801	Organic Chemical Pharmacologic Subs
cetirzine	C0055147	cetirizine	RXNORM	20610	Organic Chemical Pharmacologic Subs
Atorva 20mg	C0286651	atorvastatin 20 mg	RXNORM	83367	Organic Chemical Pharmacologic Subs
Metfomrin 500mg	C0025598	metformin	RXNORM	6809	Organic Chemical Pharmacologic Subs
Pan 40mg	C0081876	pantoprazole	RXNORM	40790	Organic Chemical Pharmacologic Subs
Paracetamol 650	C0691095	acetaminophen 650 MG Oral Tablet	RXNORM	198444	Clinical Drug

# Process Overview





# OMOP CDM Table Structure



# What We Learned

<b>De-identification</b>	Removed patient-identifiable information from clinical text
<b>Named Entity Recognition</b>	Extracted 5 main Entities : problem, test, medication, procedure, demographic details
<b>Relationship Extraction</b>	<p>25 modifiers extracted</p> <p>Examples: temporal, disease, drug strength, dosage, frequency</p> <p>Contextualized: past vs current (based on discharge summary date)</p>
<b>Concept Normalization &amp; Vocabulary Mapping</b>	<p>Mapped Indian drugs to OMOP concept IDs</p> <p>Built tool to convert extracted text/drugs to OMOP vocabulary</p>

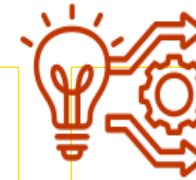
# Realities of Research



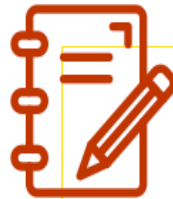
Evolving data privacy laws  
limited broader  
collaboration



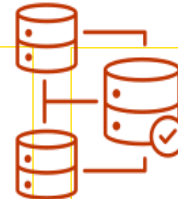
Unable to predict  
infrastructure needs for  
model development



Resource-intensive,  
managed by a lean team



Excessive time spent on  
manual data annotation



Challenges with Indian  
drug mappings and  
standardization

## Next steps

- Prepare and submit a publication on the study experience.
- Automate discharge summaries into standardized OMOP data.
- Provide an ATLAS overview and training to JSS researchers.
- Explore opportunities for networking and collaborative studies.
- Train and expand medical mapping tool for wider community use.

# Preliminary Research Questions

## **Cohort characterization:**

What are the demographic profiles, common comorbidities, procedures, drugs, and tests among inpatients with breast cancer?

## **Treatment patterns:**

Among patients with [target condition], what procedures and discharge drugs are most commonly used, and how do patterns vary by age/sex?

## **Length of Stay Prediction**

Use available features to predict prolonged hospital stays in breast cancer patients.

## **Prediction / Risk Profiling**

Can baseline demographic and clinical features predict whether a patient will complete  $\geq 6$  chemotherapy cycles?

# FHIR Work Group Initiative

Current team – Kumar Satyam (Associate Director at Providence India, HL7 Chair), Dr.Chandil (Co-founder at SVM hospital), Manish (Founder Yajur Healthcare), Parthiban (VP- Innovation & Growth, GVW), Swetha Kiranmayi (Service leader, GVW)

## Objectives

### **ABDM-FHIR to OMOP Conversion**

Enable structured mapping from ABDM-compliant FHIR to OMOP CDM for scalable, interoperable health data analytics

### **Global FHIR Guideline Adoption**

Localize international FHIR standards to align with India's digital health architecture and support OMOP integration.

### **Advance Capacity Building and Community Engagement**

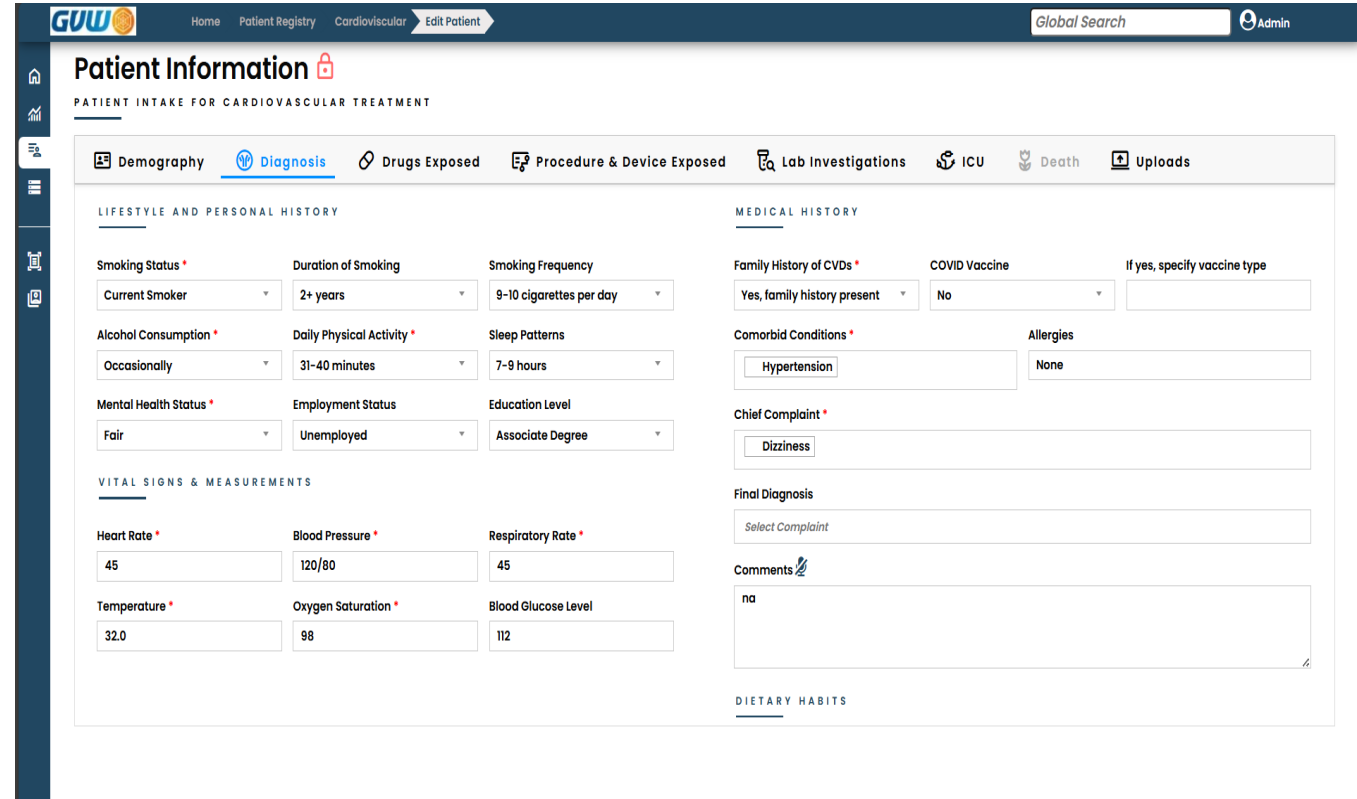
Foster knowledge exchange through workshops, documentation, and open-source contributions

# The CVD Patient Registry

**Current team** – Dr.Vikram Patil (Dy Dean,Research(Clinical), JSS AHER), Dr.Akshay (HOD Critical Care Medicine JSS Hospital), Dr.sunil Kumar (Cardiology specialist at JSS Hospital), Ranjani (Data manager at JSS Hospital), Dr. Swetha Kiranmayi (Service leader, GVW) Sushanth (Senior application engineer, GVW)

## Features –

- Enhanced data capture & gap reduction
- OMOP compatibility
- Voice-to-text input
- Predefined fields with medical dictionary integration
- Automated backend coding
- Building capabilities for ABDM-FHIR readiness



The screenshot displays the 'Patient Information' form within the GVW (Global Vascular Web) system. The form is titled 'PATIENT INTAKE FOR CARDIOVASCULAR TREATMENT' and is divided into several sections for data entry:

- Navigation Bar:** Includes links for Home, Patient Registry, Cardiovascular, and Edit Patient. A Global Search bar and an Admin link are also present.
- Form Tabs:** Demography, Diagnosis (active), Drugs Exposed, Procedure & Device Exposed, Lab Investigations, ICU, Death, and Uploads.
- LIFESTYLE AND PERSONAL HISTORY:**
  - Smoking Status: Current Smoker
  - Duration of Smoking: 2+ years
  - Smoking Frequency: 9-10 cigarettes per day
  - Alcohol Consumption: Occasionally
  - Daily Physical Activity: 31-40 minutes
  - Sleep Patterns: 7-9 hours
  - Mental Health Status: Fair
  - Employment Status: Unemployed
  - Education Level: Associate Degree
- MEDICAL HISTORY:**
  - Family History of CVDs: Yes, family history present
  - COVID Vaccine: No
  - If yes, specify vaccine type: (empty field)
  - Comorbid Conditions: Hypertension
  - Allergies: None
  - Chief Complaint: Dizziness
  - Final Diagnosis: Select Complaint
  - Comments: na
- VITAL SIGNS & MEASUREMENTS:**
  - Heart Rate: 45
  - Blood Pressure: 120/80
  - Respiratory Rate: 45
  - Temperature: 32.0
  - Oxygen Saturation: 98
  - Blood Glucose Level: 112
- DIETARY HABITS:** (Section header visible at the bottom)

# Digital Health Centre of Excellence

**Current team** – Rintu Kutum (Group leader, Augmented Health Systems Laboratory. Faculty Fellow of Computer Science. Koita Centre for Digital Health at Ashoka. Mphasis AI & Applied Tech Lab at Ashoka. Trivedi School of Biosciences. Ashoka University), Parthiban (VP-Innovation & Growth, GVW), Swetha Kiranmayi (Service leader, GVW)

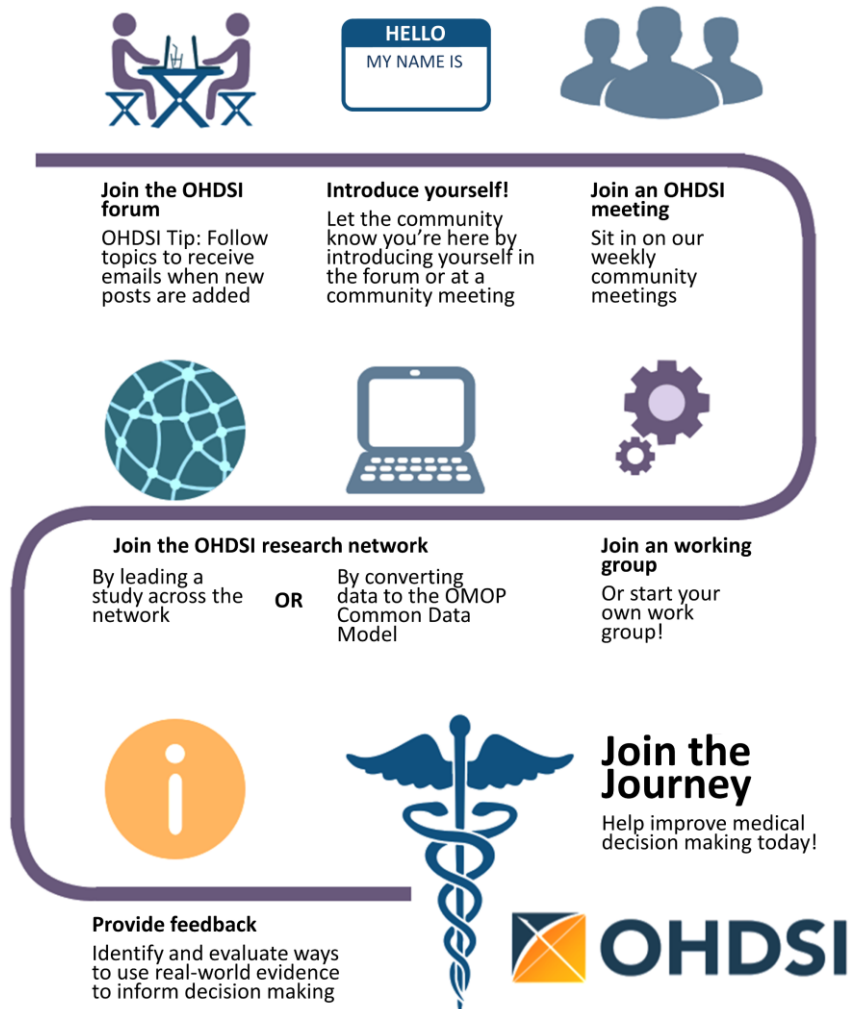
**Aim** – To unlock the potential of large-scale health data in India by enabling real-world data (RWD) generation

## Objectives

- Establish a collaborative hub for digital health innovation, standardization, and capacity-building.
- Drive ABDM & FHIR-aligned digital health technologies.
- Engage experts and stakeholders across clinical, technical, and academic domains.
- Use impactful case studies to inform policy and best practices.
- Build an integrated ecosystem combining data, technology, and evidence.
- Promote open science by linking FHIR & OMOP for reproducible research.
- Advance semantic interoperability for Indian health data integration and reuse.
- Secure funding to sustain and scale OHDSI India initiatives.



# Join the journey - How to become an OHDSI collaborator



Register here: [Be a Collaborator](#)

