# **Development of Computable Operational Definitions to Maximize Comparability & Consistency Across a Multi-Data Source Global Real-World Effectiveness Program**

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#### Disclosures

• A.K., C.P., and A.S. are partners and owners of Navidence LLC, which is subcontracted via Ikaika Health LLC to provide support to AstraZeneca for various projects and studies • L.G., S.D., C.T., and S.T. are employees of AstraZeneca

# Introduction

- AZD7442, a combination of monoclonal antibodies (tixagevimab/cilgavimab), received emergency use authorization by the FDA in December 2021 for pre-exposure prophylaxis (PrEP) against COVID-19 in patients who are moderately to severely immunocompromised (IC)
- Essential real-world effectiveness assessment is challenged by heterogeneous definitions of IC eligibility, drug distribution differences, inconsistent clinical practices, record-keeping, and availability and completeness of key data across geographic regions/countries and data sources

# **Objective**

• Develop data source-agnostic computable operational definitions (cODs) to support a multinational, multi-data source real-world effectiveness program

# **Key Take-Home Message**

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Computable operational definitions support a *multi-data source global* real-world effectiveness program:

- Direct clear and consistent queries across data sources
- Standards-based code lists/value sets
- References for justification of operational definitions and code list mappings (specific to the study element type)
- Context for comparison and interpretation of results across and between data sources

### Conclusions

- Every data source is distinct within and across geographic regions/countries, and the unique context is important to the interpretation of results
- Data source-agnostic cODs are foundational to maximize consistency across data sources and countries, comparability of study results, and support reproducibility, even when context may vary
- The authors advise thoughtful creation of clear, consistent, standards-based cODs for all real-world evidence studies, especially those using multiple data sources and/or those submitting evidence to external stakeholders (e.g., regulatory agencies)

## Methods

#### 1. Data source-agnostic cODs were established for each of the study elements of the umbrella protocol: eligibility criteria, exposure, baseline characteristics, and outcome measures



- 3. cODs were tailored to reflect differences for each data source
- Selected data sources are completing the analysis independently of each other to ensure context and uniqueness of the underlying health system are accounted for

#### Figure. Leveraging data source-agnostic cODs to create data source-specific cODs



Outcome measures cODs

	Components of a cOD:	≥ 1 Medication Record from Cancer Therapies within the 6 months prior to	Any Malignancy, except malignant neoplasm of skin (CCI) Diagnoses (SNOMED)	extensional 716	Data source C ——
	Data variable	the index date	Metastatic Solid Tumor (CCI) Diagnoses Metastatic Solid Tumor (CCI) Diagnoses (ICD-10-CM)	intensional 69	Outcome measures cC
	Quantity/magnitude		Metastatic Solid Tumor (CCI) Diagnoses (ICD-9-CM)	intensional 0	
	<ul> <li>Target value, concept, or code list/value set</li> </ul>		Metastatic Solid Tumor (CCI) Diagnoses (SNOMED)	extensional 51	4. Development of cODs was an iterative process and included:
	Time period qualifier		Cancer Therapies	grouping 0	<ul> <li>review of literature and published protocols</li> </ul>
	• Other qualifiers based on the data variable or type of operational definition		Cancer Therapies (ATC)	extensional 263	<ul> <li>input from epidemiology, statistics, informatics, and medical teams</li> </ul>
	Expression to describe how the components relate to each other		Cancer Therapies (Name)	extensional 263	<ul> <li>– clinical concepts represented by commonly structured data types and standard validated code lists</li> </ul>
N N			Cancer Therapies (RxNorm)	extensional 258	

# **Results and interpretation**

- cODs for 174 study elements were developed for the umbrella study protocol:
- 17 eligibility criteria
- 1 exposure
- 68 baseline characteristics
- 88 outcomes (for 16 objectives)

# Limitations

- This program was conducted within the context of a large global real-world effectiveness program covering several data sources across the US, Israel, and specific countries in the EU. Therefore, it does not include countries across all regions of the world, which may limit generalizability of results
- However, the authors suggest that these methods of using clear, consistent, standards-based cODs would provide greater value for programs conducted across more regions of the world

#### · The cODs encompass 38 distinct data variables and 82 standards-based code lists (further delineated by a code system)

Label	Code	Essential
Age	age	True
Date of Birth	date_of_birth	True
Death Record	death_record	True
Encounter Disposition	encounter_disposition	True
Encounter Record	encounter_record	True
Enrollment End Date	enrollment_end_date	True
Enrollment Start Date	enrollment_start_date	True
Sex	sex	True
T-cell helper (CD4) subset panel (Bld)	65758-5	True
Vaccination Record	vaccination_record	True
Admission Date	admission_date	False
Alcohol Use	alcohol_use	False
Area deprivation index	area_deprivation_index	False
Body Mass Index	body_mass_index	False
Cancer Screening Encounter Record	cancer_screening_encounter_record	False
COVID-19 Lab Tests	covid-19_lab_tests	False
Diagnosis Record	diagnosis_record	False
Discharge Date	discharge_date	False
Eligible Dependents	eligible_dependents	False
Immunization record	immunization_record	False
Insurance type	insurance_type	False
Isolation event	isolation_event	False
Marital Status	marital_status	False
Median household salary	median_household_salary	False
Medication Record	medication_record	False

#### • Direct comparisons across data sources were demonstrated to highlight differences in cODs and their potential impact on the interpretation of analysis results



- While this program resulted in the creation of cODs that cover a variety of medical conditions (either as immunosuppressive conditions for eligibility, many medical conditions as possible comorbidities, or as select effectiveness or safety outcomes), this did not include all disease therapy areas, nor the distinct subtypes of all related medical conditions
- The authors advise thoughtful creation of clear, consistent, standards-based cODs for medical conditions that are relevant to individual clinical research needs and purpose
- In addition, the authors advise leveraging cODs with the appropriate context of the study element. For example, different cODs may be appropriate for the same clinical condition/conceptual definition, when applied as an inclusion criterion versus a baseline comorbidity versus an outcome measure

#### Supplementary content



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- All data sources required some adaptation, primarily on coding schemas or definitions unique to the data source
- Examples of coding schemes include:

• Diagnosis codes: ICD-10-CM, ICD-9, SNOMED

- Procedure codes: ICD-10-PCS, ICD-9, CPT, HCPCS
- Medication codes: RxNorm, ATC-5 (as well as generic and trade names)
- Examples of unique definitions between data sources:
- IC alignment with national eligibility requirements (e.g., PrEP vs Treatment, IC definitions)
- Setting of AZD7442 administration (e.g., inpatient vs outpatient; IC speciality care vs infectious disease department)
- Patient/population characteristics: socioeconomic status, geographic distribution, common healthcare practices

#### **Abbreviations and codes**

ATC-5, Anatomical Therapeutic Chemical 5th level; CCI, Charlson Comorbidity Index; CM, Clinical Modification; cOD, computable operational definition; COVID-19, coronavirus disease 2019; CPT, Current Procedural Terminology; EUA, Emergency Use Authorization; FDA, The US Food and Drug Administration HCPCS, Healthcare Common Procedure Coding System; IC, immunocompromised; ICD-9, International Classification of Diseases, 9th Revision; ICD-10, International Classification of Diseases, 10th Revision; LOINC, Logical Observation Identifiers Names and Codes; PCR, polymerase chain reaction; PCS, Procedure Coding System; PrEP, pre-exposure prophylaxis; RxNorm, standardized nomenclature for clinical drugs; SARS-CoV-2, severe acute respiratory syndrome coronavirus 2; SNOMED, Systematized Nomenclature of Medicine.

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