

— — OHDSI

Observational Health Data Sciences and Informatics (OHDSI)

2025-03-02

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OHDSI OHDSI OHDSI
OHDSI OHDSI OHDSI

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- I. OHDSI□□□□□□
 - II. □□□□□□□□□□
 - III. □□□□□
 - IV. □□□□□□□
 - V. OHDSI□□

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Clair Blacketer	David Blatt	Brian Christian
Gino Cloft	Frank DeFalco	Sara Dempster
Jon Duke	Sergio Eslava	Clark Evans
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Bob Lanese	Wanda Lattimore	Chun Li
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Jamie Weaver	James Wiggins	Andrew Williams
Seng Chan You		

OHDSI

- ACHILLES: 1.6.6
- ATLAS: 2.7.3
- EUNOMIA: 1.0.0
- OMOP: 1.0.0

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RMarkdown bookdown
<https://github.com/OHDSI/TheBookOfOhdsiInJapanese/>
 “travis”

Table 1: Methods Library

CaseControl	1.6.0
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SelfControlledCohort	1.5.0
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Part I

OHDSI

Chapter 1

— — OHDSI

□□□□□□□□□□□□□□□□□□ & □□□□□□□□□□□□

1.1

1.2

Observational Medical Outcomes Partnership OMOP OMOP FDA (Stang et al., 2010) OMOP

ОМОН

CMC1 CMC2 CMC3 CMC4 CMC5 CMC6 CMC7 CMC8 CMC9 CMC10 CMC11 CMC12 CMC13 CMC14 CMC15 CMC16 CMC17 CMC18 CMC19 CMC20 CMC21 CMC22 CMC23 CMC24 CMC25 CMC26 CMC27 CMC28 CMC29 CMC30 CMC31 CMC32 CMC33 CMC34 CMC35 CMC36 CMC37 CMC38 CMC39 CMC40 CMC41 CMC42 CMC43 CMC44 CMC45 CMC46 CMC47 CMC48 CMC49 CMC50 CMC51 CMC52 CMC53 CMC54 CMC55 CMC56 CMC57 CMC58 CMC59 CMC60 CMC61 CMC62 CMC63 CMC64 CMC65 CMC66 CMC67 CMC68 CMC69 CMC70 CMC71 CMC72 CMC73 CMC74 CMC75 CMC76 CMC77 CMC78 CMC79 CMC80 CMC81 CMC82 CMC83 CMC84 CMC85 CMC86 CMC87 CMC88 CMC89 CMC90 CMC91 CMC92 CMC93 CMC94 CMC95 CMC96 CMC97 CMC98 CMC99 CMC999

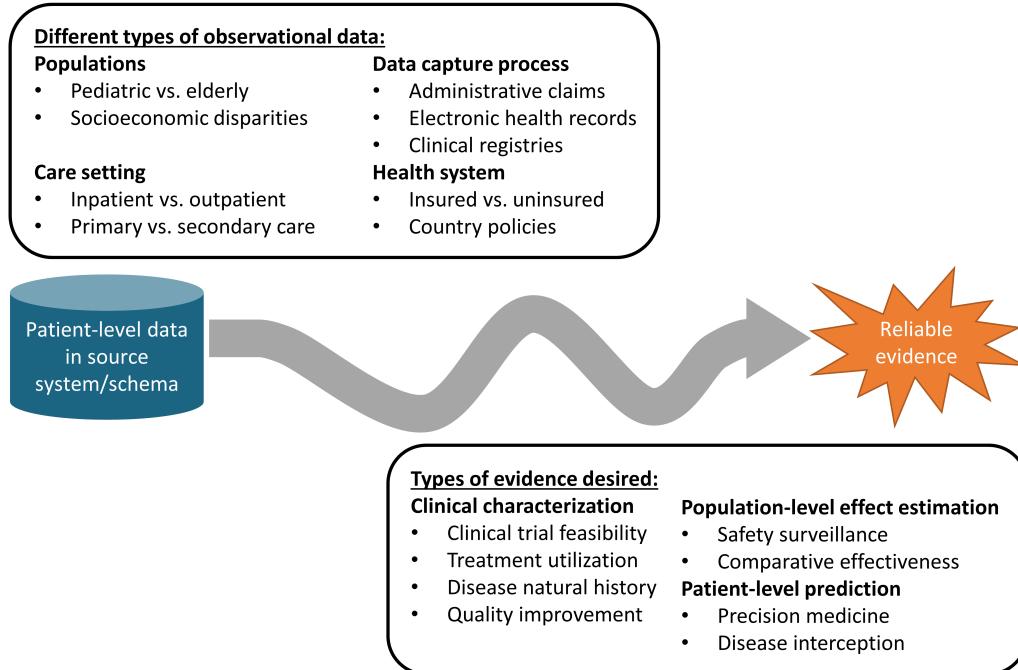
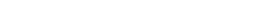


Figure 1.1: □□□□□□□□□□□□

OMOP (Ryan et al., 2012, 2013b) (Schuemie et al., 2014) (Madigan et al., 2013b,a) OMOP (OMOP-CC) (OMOP-CC)

OMOP



- ETL -
 -
 -

OHDSI

OHDSI

Observational Health Data Sciences and Informatics (OHDSI) (et al., 2015) (OHDSI)

1.2.1

1.2.2



1.2.3

1.3 OHDSI

OHDSI 2014 OHDSI 1.2



Figure 1.2: 2019 8 OHDSI

2019-08-00 OHDSI 20-00-00 100 CDM OHDSI

¹<https://www.ohdsi.org/who-we-are/collaborators/>

OHDSI OMOP CDM 2) OHDSI OMOP CDM OHDSI Academy of Science 5000 11 OHDSI (Tian et al., 2018) et al., 2018a) (Duke et al., 2017) (Vashisht et al., 2018) et al., 2018b) OHDSI et al., 2018) (Johnston et al., 2019; Cepeda et al., 2018; Reps et al., 2019)

1.4 OHDSI

OHDSI OHDSI 2

1.5



- OHDSI
- OHDSI
- OHDSI

²<https://github.com/OHDSI>

Chapter 2

OHDSI & OMOP

OHDSI - OMOP

OHDSI

2.1

OHDSI

2.1.1 OHDSI

OHDSI¹ OHDSI

OHDSI

- : OHDSI
- : OHDSI
- : OHDSI OMOP CDM
- : CDM OHDSI
- : CDM OHDSI
- : OMOP OHDSI

to OHDSI! - Please introduce yourself OHDSI - OHDSI- OHDSI (2)

¹<https://forums.ohdsi.org>



2.1.2 OHDSI

OHDSI

2.1.3 OHDSI

OHDSI

OHDSI □□□□□□□□□□□□□□□□□□ OHDSI □□□□□□□□□□□□□□□□□□

2.1.4 OHDSI

OHDSI

Wiki OHDSI Wiki](<https://www.ohdsi.org/web/wiki/doku.php?id=projects:overview>)

□ □ □ □ □ OHDSI □ □ □ □ □

□ □ □ □ □ □ □ □

JavaScript

CDM & Vocabulary

□□□□□□□□□□□□□□□ OMOP □□□□□□□□□□□□□□□ OMOP

Genomics

OMOP CD- 

CDM

Population-Level Estimation

2.1.5 OHDSI

2.1.6 OHDSI

OHDSI OMOP OHD
CDM

2.2

²<https://www.ohdsi-europe.org/>

³<https://forums.ohdsi.org/c/For-collaborators-wishing-to-communicate-in-Korean>

⁴<https://ohdsichina.org/>

2.3



- OHDSI
 - ETL OHDSI

Chapter 3

□ □ □ □ □ : □ □ □ □ □ □ □ □ □ □ □ □

OHDSI

OHDSI

OHDSI

3.1

1990 20
(Wikipedia, 2019a)

3.2 : the Study-a-Thon

study-a-thon OMOP a-thon 2 study-a-thon OHDSI Chapter 8 ATLAS OHDSI

3.3

3.4

OHDSI Methods Library 2020 GitHub Library

3.5

//howoften.org □ http://data.ohdsi.org □ 20□ OMOP
CDM □ OHDSI ETL

3.6

[wiki](#) [GitHub](#) [OHDSI](#)

3.7 OHDSI FAIR

3.7.1

Wilkinson et al. (2016) FAIR OHD-SI

3.7.2

OMOP 2019 IMI EHDEN

3.7.3

OMOP CD-
SQL OMOP
EHDEN LEGEND <http://how>

3.7.4

□ OMOP □ OHDSI □ FHIR □ HL7 □ CIMI □ openEHR □ OHDSI □ Athena □ OHDSI □ OMOP

3.7.5

□ FAIR □ ETL □ CDM



- OHDSI □ OMOP □ OHDSI □ FHIR □ HL7 □ CIMI □ openEHR □ OHDSI □ Athena □ OHDSI □ OMOP
- □ FAIR □ ETL □ CDM

4 — —

Chapter 4

□ □ □ □: *Clair Blacketer*

A horizontal row of 20 empty square boxes, intended for students to write their answers in a grid format.

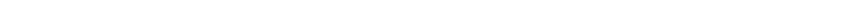
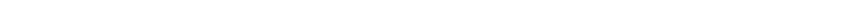
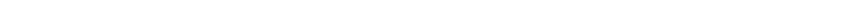
CDM 5 CDM

CDM 4.1

4.1

CDM

CDM

- **CDM**: 
 - 
 - 

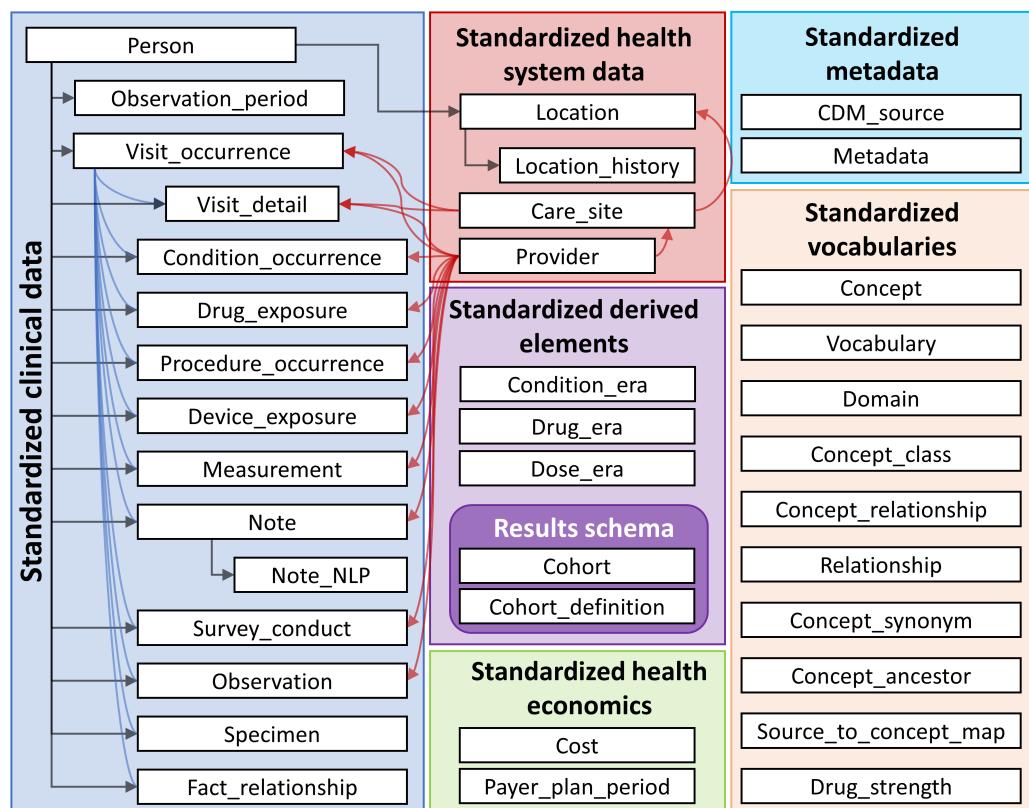


Figure 4.1: CDM 6.0

4.2

4.2.1

4.2.2

10 COHORT

4.2.3

4.2.4

□□□□□□	□□□□ID	□□□□□□	□□□□ID
1731378	Drug	183	Route
477597	Device	180	Currency
257000	Procedure	158	Payer
163807	Condition	123	Visit

□□□□□	□□□□ID	□□□□□	□□□□ID
145898	Observation	51	Cost
89645	Measurement	50	Race
33759	Spec Anatomic Site	13	Plan Stop Reason
17302	Meas Value	11	Plan
1799	Specimen	6	Episode
1215	Provider Specialty	6	Sponsor
1046	Unit	5	Meas Value Operator
944	Metadata	3	Spec Disease Status
538	Revenue Code	2	Gender
336	Type Concept	2	Ethnicity
194	Relationship	1	Observation Type

4.2.5

4.2.6

4.2.7

- ICD9CM □ NDC □ Read □
 - CDM □

- CDM
- CDM

QA
4.2 ICD9CM 011



The screenshot shows the ATHENA interface with the title "Pulmonary tuberculosis". Below it is a table titled "DETAILS" containing the following information:

Domain ID	Condition
Concept Class ID	3-dig nonbill code
Vocabulary ID	ICD9CM
Concept ID	44828631
Concept code	011
Invalid reason	Valid
Standard concept	Non-standard
Synonyms	Pulmonary tuberculosis
Valid start	12/31/1969
Valid end	12/30/2099

Figure 4.2: ICD9CM

011 UB04 A DRG
TB Figure 4.3 OMOP
253954 Read ICD10 CIEL MeSH

4.7

4.3 CDM

CDM 16 10 2 4
CDM Wiki¹

1 https://github.com/OHDSI/CommonDataModel/wiki

TERM CONNECTIONS (82)			
RELATIONSHIP	RELATES TO	CONCEPT ID	VOCABULARY
ICD-9-CM to MedDRA (MSSO)	Pulmonary tuberculosis	36110777	MedDRA
Non-standard to Standard map (OMOP)	Pulmonary tuberculosis	253954	SNOMED
Subsumes	Other specified pulmonary tuberculosis	44830894	ICD9CM
	Other specified pulmonary tuberculosis, bacteriological or histological examination not done	44836741	ICD9CM
	Other specified pulmonary tuberculosis, bacteriological or histological examination unknown (at present)	44836742	ICD9CM
	Other specified pulmonary tuberculosis, tubercle bacilli found (in sputum) by microscopy	44821641	ICD9CM
	Other specified pulmonary tuberculosis, tubercle bacilli not found (in sputum) by microscopy, but found by bacterial culture	44833188	ICD9CM

Figure 4.3: □□□□SNOMED□□□

4.3.1 :



Lauren 
[//endometriosis-uk.org/laurens-story](http://endometriosis-uk.org/laurens-story) 

4.3.2 PERSON

- □□□36□□□□
 - □□□□□□□1982□3□12□□
 - □□□□□□□
 - □□□□□□□□□□□

PERSON

Table 4.3: PERSON□□□□□

PERSON_ID	1	PERSON_ID
GENDER_CONCEPT_ID	8532	ID 8532
YEAR_OF_BIRTH	1982	
MONTH_OF_BIRTH	3	
DAY_OF_BIRTH	12	
BIRTH_DATETIME	1982-03-12 00:00:00	
DEATH_DATETIME		
RACE_CONCEPT_ID	8527	ID 8527
ETHNICITY_CONCEPT_ID	38003564	US
LOCATION_ID		
PROVIDER_ID		
CARE_SITE		
PERSON_SOURCE_VALUE	1	
GENDER_SOURCE_VALUE	F	
GENDER_SOURCE_CONCEPT_ID	0	OHDSI
RACE_SOURCE_VALUE	white	PCORNet
RACE_SOURCE_CONCEPT_ID	0	
ETHNICITY_SOURCE_VALUE	english	GENDER_SOURCE_CONCEPT_ID
ETHNICITY_SOURCE_CONCEPT_ID	0	GENDER_SOURCE_CONCEPT_ID

4.3.3 OBSERVATION_PERIOD

Table 4.4: Lauren□□□□□□□□□□□□

□□□□□□□ID	□□□	□□□	□□□
70	2010-01-06	2010-01-06	□□□□

ID			
80	2011-01-06	2011-01-06	
90	2012-01-06	2012-01-06	
100	2013-01-07	2013-01-07	
101	2013-01-14	2013-01-14	
102	2013-01-17	2013-01-24	

OBSERVATION_PERIOD

4.3.4 VISIT_OCCURRENCE

VISIT_OCCURRENCE OHDSI

VISIT_OCCURRENCE

Table 4.6: VISIT_OCCURRENCE

VISIT_OCCURRENCE_ID	514	
PERSON_ID	1	PERSON Laura
VISIT_CONCEPT_ID	9201	9201
VISIT_START_DATE	2013-01-17	
VISIT_START_DATETIME	2013-01-17 00:00:00	
VISIT_END_DATE	2013-01-24	
VISIT_END_DATETIME	2013-01-24 00:00:00	

VISIT_TYPE_	32034	EHR
CONCEPT_ID		
PROVIDER_ID	NULL	
CARE_SITE_ID	NULL	
VISIT_SOURCE_	□□	
VALUE		
VISIT_SOURCE_	NULL	OHDSI
CONCEPT_ID		
ADMITTED_FROM_	NULL	
CONCEPT_ID		8536
ADMITTED_FROM_	NULL	
SOURCE_CONCEPT_ID		
DISCHARGE_TO_	NULL	
CONCEPT_ID		8615
DISCHARGE_TO_	NULL	
SOURCE_VALUE		living facility
PRECEDING_VIS		

IT_OCCURRENCE_ID|NULL|ADMITTED_FROM_CONCEPT

-  wiki 

4.3.5 CONDITION_OCCURRENCE

CONDITION OCCURRENCE

Table 4.7: CONDITION_OCCURRENCE□□□□□

EHR ID 32020 |
|CONDITION_STATUS_CONCEPT_ID|NULL|
4203942 |STOP_REASON|NULL|

4.4

4.5



4.6

1

CDM ATHENA³ ATLA

²<https://github.com/OHDSI/CommonDataModel/wiki>

³<http://athena.ohdsi.org/>

⁴<http://atlas-demo.ohdsi.org>

□□ 4.3. □□□□ 2019 □ 5 □ 1 □□□□□□□□□ 200 MG □□□□□ NDC □□□□□ 76168009520 □□ 30 □

1

□□□□□3□□□□□□□□□□□□□□□@ref□installIR□□□□□□□□□□□R□R-Studio□□□□Java□□

```
install.packages(c("SqlRender", "DatabaseConnector", "remotes"))
remotes::install_github("ohdsi/Eunomia", ref = "v1.0.0")
```

Eunomia R CDM

```
connectionDetails <- Eunomia::getEunomiaConnectionDetails()
```

CDM main CONDITION OCCURRENCE

```
library(DatabaseConnector)
connection <- connect(connectionDetails)
sql <- "SELECT *
FROM @cdm.condition_occurrence
LIMIT 1;"
result <- renderTranslateQuerySql(connection, sql, cdm = "main")
```

□ 4.4. SOLR ID192671

□□ 4.6. SQL R □□□□□ PERSON ID 61 □□□□□□□□□□□□□□□□□□□□□□□□□

□□□□□□□□□□@ref□Cdmanswers□□□□□□□

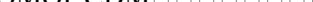
Chapter 5

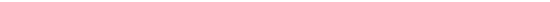
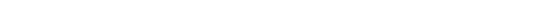
5 — —

□□□□: Christian Reich & Anna Ostropolets

5.1

of Mortality 5.1

OHDSI  OMOP CDM  CDM 

- 
 - 

5.1.1

CDM OHDSI

GitHub 3

¹<https://github.com/OHDSI/Vocabulary-v5.0>

²<https://forums.ohdsi.org>

³<https://github.com/OHDSI/CommonDataModel/issues>

1669.

A General BILL for this present Year,

Ending the 11th Day of December 1660.

According to the Report made to the King's most excellent Majesty,
By the Company of Parish Clerks of LONDON, &c.

DISEASES and CASUALTIES.

A	Bortive and Stillborn	421	Flox and Small Pox	—	1523	Palsy	—	—	17
	Aged	909	Found dead in the Streets,	{		Plague	—	—	36
	Ague and Fever	2303	Fields, &c.	2		Plurify	—	—	12
	Apoplexy and Suddenly	91	French Pox	—	51	Quinny and fore Throat	—	—	21
	Blafted and Planet	3	Gout	—	4	Rickets	—	—	441
	Bleeding and bloody Issue	7	Grief	—	13	Rising of the Lights	—	—	210
	Bloody Flux, Scowring, and Flux	346	Griping in the Guts	—	253	Rupture	—	—	12
	Burnt and Scalded	6	Hanged and made away themselves	{	11	Scurvy	—	—	82
	Cancer, Gangrene and Fistula	63	Head-ach and Headmouldshot	—	35	Shot	—	—	7
	Canker, fore Mouth and Thrush	73	Jaundies	—	102	Shingles	—	—	1
	Childbed	226	Imposthume	—	105	Sores, Ulcers, broken and bruised Limbs	{	—	61
	Chrisomes and Infants	858	Killed by several Accidents	—	55	Spleen	—	—	7
	Cold, Cough and Hiccup	33	King's Evil	—	28	Spotted Fever and Purples	—	—	368
	Colick and Wind	116	Lethargy	—	6	Starved	—	—	7
	Consumption and Tiswick	2982	Livergrown	—	8	Strangury	—	—	22
	Convulsion	742	Lunatick and Frenzy	—	14	Stopping of the Stomach	—	—	186
	Cut of the Stone and Stone	46	Megrims	—	5	Surfeit	—	—	202
	Dropfy and Tympany	646	Measles	—	6	Swine Pox	—	—	2
	Drowned	57	Mother	—	7	Teeth and Worms	—	—	839
	Executed	7	Murthered	—	7	Vomiting	—	—	8
	Falling Sicknes	4	Overlaid and Starved at Nurse	—	46	Wen	—	—	1

Figure 5.1: 1660 62

5.1.2

□□□□□□□□□□□□□□□□□□□□□□□□□ Pallas □□□□□□□□□□□□□ ATHEN

OMOP CDM zip

5.1.3 :

OHDSI Extension 5.6.9

5.2

OMOP CDM

CONCEPT_ID	313217	Primary key
CONCEPT_NAME	Atrial fibrillation	English description
DOMAIN_ID	Condition	Domain
VOCABULARY_ID	SNOMED	Vocabulary
CONCEPT_CLASS_ID	Clinical Finding	Class in vocabulary
STANDARD_CONCEPT	S	Standard, Source of Classification
CONCEPT_CODE	49436004	Code in vocabulary
VALID_START_DATE	01-Jan-1970	Valid during time interval
VALID_END_DATE	31-Dec-2099	
INVALID_REASON		

Figure 5.2: OMOP CDM

5.2.1 ID

5.2.2

⁴<http://athena.ohdsi.org>

5.2.3

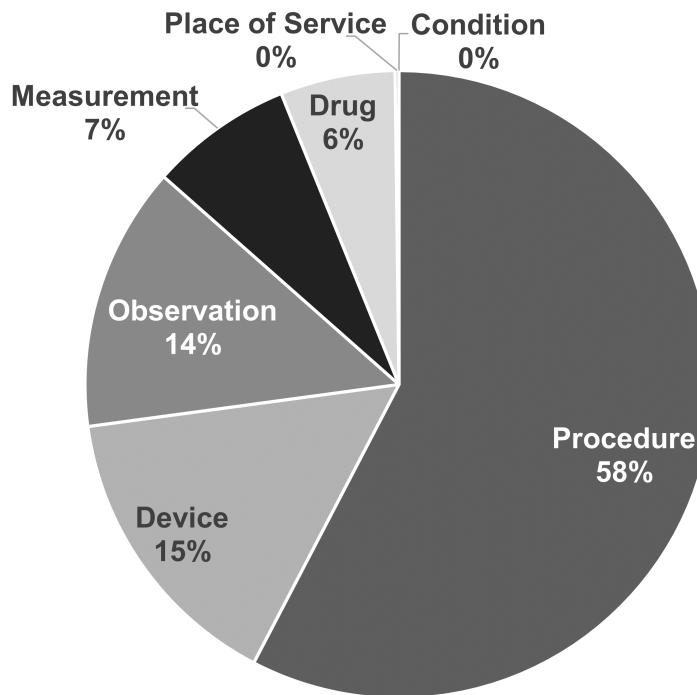


Figure 5.3: CPT4 HCPCS

4 **5.6** CDM

5.2.4

5.2.5

Class ID

A horizontal row of 15 empty rectangular boxes, likely used for input fields or placeholder text in a form.

RxNox

5.2.6

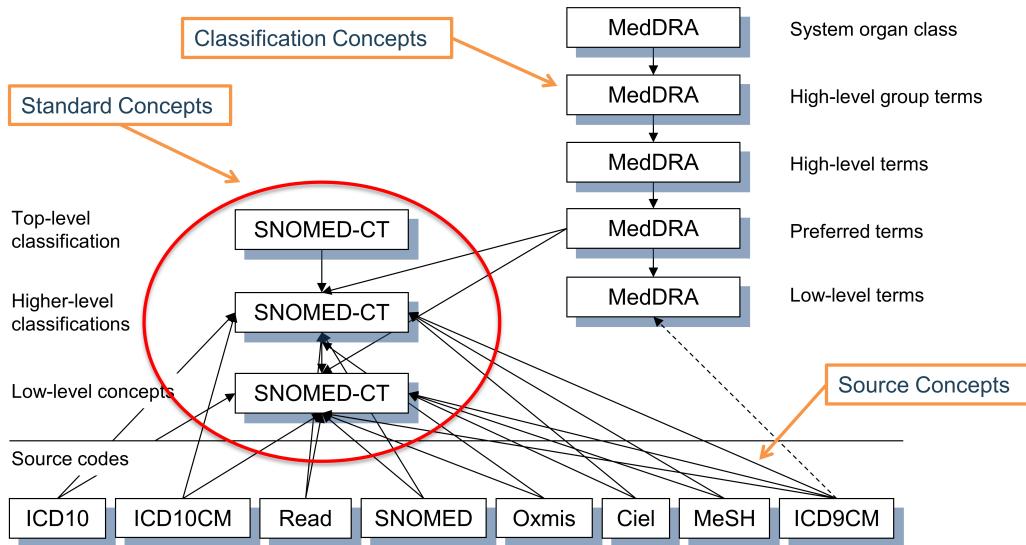
5.2.7

5,2,8

- □ 5.4 □ □ □

5.2.9

ICD9CM NDC



ID		ID	ID	
35803438	1001	□□□□□□□□□□□□	HemOnc	□□□□□□□□□□□□
35942070	1001	AJCC TNM Clin T	□□	NAACCR NAACCR□□
1036059	1001	□□□□□□□	□□	RxNorm
38003544	1001	□□□□□□□□□□□□	□□□□□□	□□□□□□
		- □□□		
43228317	1001	□□□□□□□□□□□□□□	BDPM	□□
45417187	1001	□□□□□□□□□□□□□□	Multum mg/ml□□□□□	Multum
45912144	1001	□□	□□	CIEL

5.2.10

5.3

□□□2□□□□□□□□□□□□□□□□□
to□□□□□□□□□□□ Mapped from□□□□□

5.3.1

Table 5.4: □□□□□□□□□□□

“Maps to”
“Maps to value”
“mapped from”

“Mapped from”

MEASUREMENT
OBSERVATION
VALUE_AS_CONCEPT

W61.51 SNOMED
217716004

ICD9CM
070.43 E SNOMED 235867002 E SNOMED
72836002 2

- □□□□□□□□□□□□
 - □□□□□□□□□□□□
 - □□□□□□□□□□□□
 - □□□□□□□□□□□□

to Maps to value

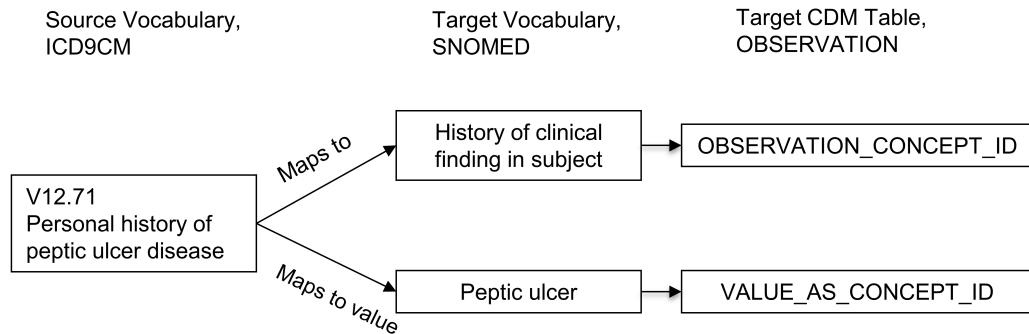


Figure 5.5: to Maps to value

OMOP OHDSI Wiki⁵

5.3.2

Is a - Subsumes 49436004 SNOMED 17366009 Is a 49436004 SNOMED 40593004 Is a

5.3.3

Vocabulary A - Vocabulary B equivalent - RxNorm equivalent Maps to

5.3.4

OHDSI Wiki⁶

site of 5.5

⁵<https://www.ohdsi.org/web/wiki/doku.php?id=documentation:vocabulary:mapping>

⁶<https://www.ohdsi.org/web/wiki/doku.php?id=documentation:vocabulary>

Table 5.5: Finding site

CONCEPT_ID_1	CONCEPT_ID_2
4000504 “Urethra part”	36713433 “□□□□□□□”
4000504 “Urethra part”	433583 “□□□□□□”
4000504 “Urethra part”	443533 “□□□□□□□□”
4000504 “Urethra part”	4005956 “□□□□□□□□”

5.4

CONCEPT_ANCESTOR CONCEPT_RELATIONSHIP
“Is a” - “Subsumes” 5.6

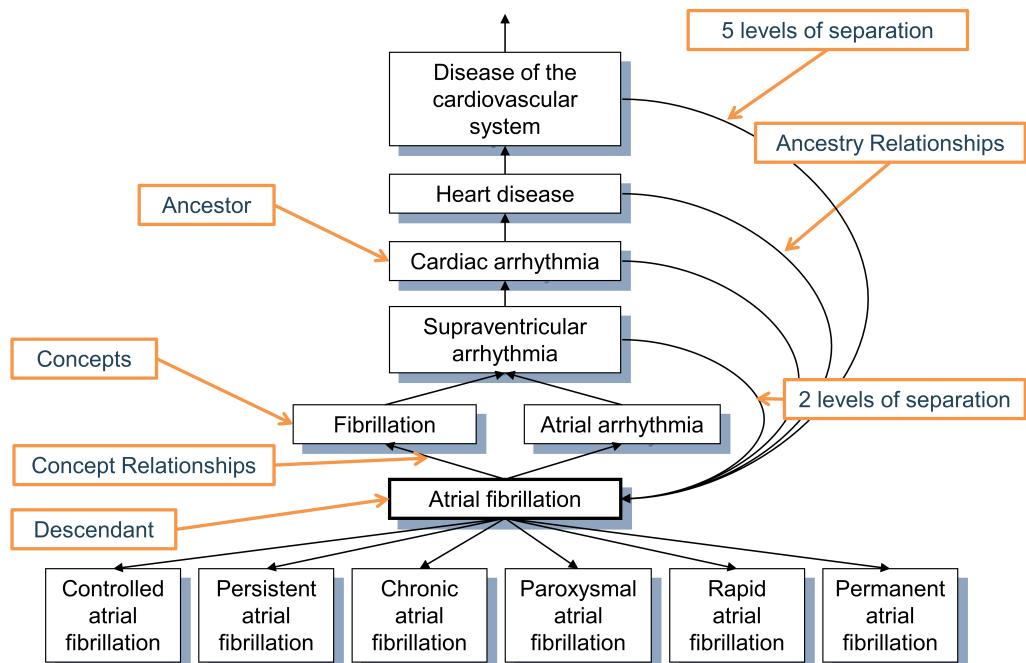


Figure 5.6: Is a Subsumes

5.5

5.6

5.6.1

OMOP CDM

5.6.2

5.6.3

0MOP

ICD-9 ICD-10

5.6.4

HCPCS CPT4

5.6.5

5-6-6

5.6.7

5.6.8

5_6_9

RxNorm Extension

5.6.10 NULL

5 8507 8

5.7



- OMOP
 - OHD
 - CDM
 - Mapping
 - Maps to value
 -
 -
 -

5.8

10 / 10

□ □ 5.1. “□ □ □ □” □ □ □ □ □ □ ID □ □ □ □ □

5.3. “ ” MedDRA

□□□□□□□□□ E.2 □□□□□□□□□

⁷<http://athena.ohdsi.org/>

⁸<http://atlas-demo.ohdsi.org>

Chapter 6

6 — —

□□□□: *Clair Blacketer & Erica Voss*

6.1

OMOP / CDM ETL

ETL 4

1. CDM
 2.
 3. ETL
 4.

OHPSI

□□□□ 1· ETL□□□□□

6.1.1 White Rabbit

Rabbit ETL White Rabbit OMOP CDM ETL Rabbit ETL

¹<https://github.com/OHDSI/WhiteRabbit>

□ □ □ □ □

ANSWER

A horizontal row of 20 empty square boxes for writing responses.

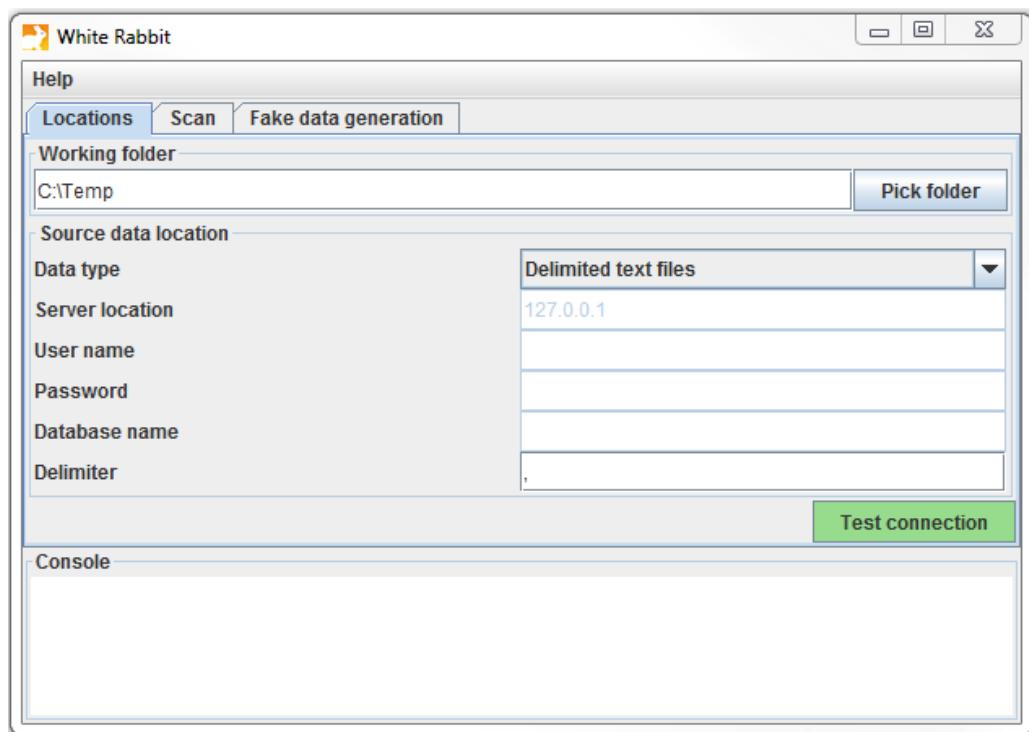


Figure 6.1: White Rabbit Pick
Folder

White Rabbit

A horizontal row of fifteen empty square boxes, intended for children to draw or write in.

ETL + all in DB

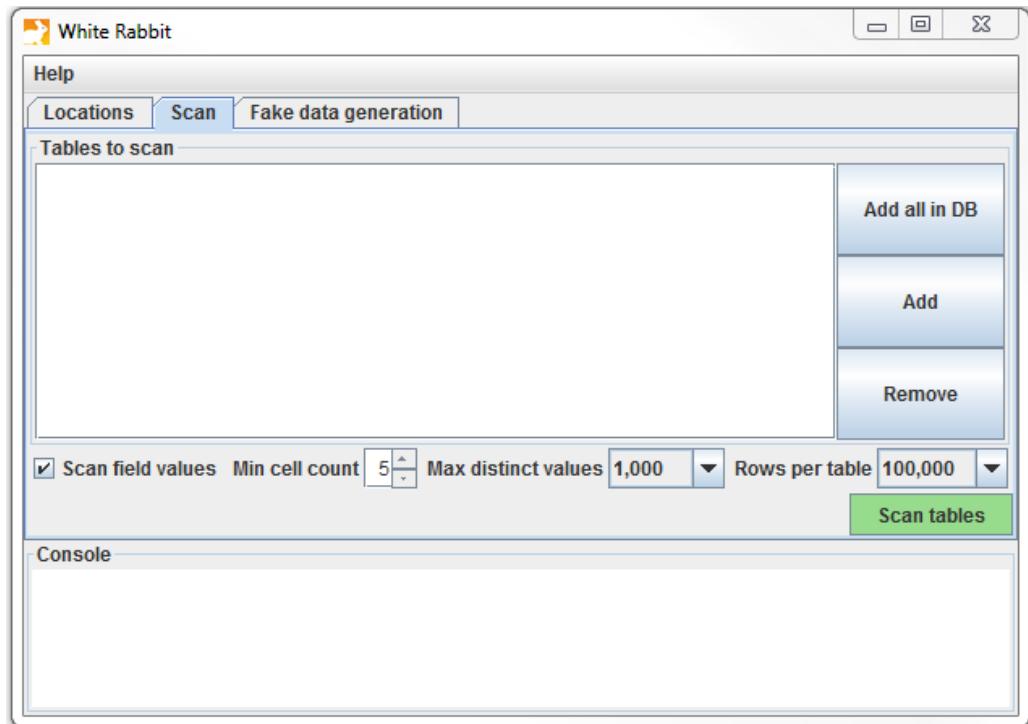


Figure 6.2: White Rabbit □□□□□□□□

A horizontal row of 20 empty rectangular boxes, likely for students to write their answers in a worksheet.

- White Rabbit
 - 5
 - White Rabbit 100,000

Excel

A	B	C	D	E	F	G
1 Table	Field	Type	Max length	N rows	N rows checked	Fraction empty
2 dbo.allergies	start	date	10	3184	3184	0
3 dbo.allergies	stop	date	10	3184	3184	0.725188442
4 dbo.allergies	patient	varchar	36	3184	3184	0
5 dbo.allergies	encounter	varchar	36	3184	3184	0
6 dbo.allergies	code	varchar	9	3184	3184	0
7 dbo.allergies	description	varchar	24	3184	3184	0
8						
9 dbo.careplans	id	varchar	36	30199	30199	0
10 dbo.careplans	start	date	10	30199	30199	0
11 dbo.careplans	stop	date	10	30199	30199	0.057849598
12 dbo.careplans	patient	varchar	36	30199	30199	0
13 dbo.careplans	encounter	varchar	36	30199	30199	0
14 dbo.careplans	code	varchar	15	30199	30199	0
15 dbo.careplans	description	varchar	62	30199	30199	0
16 dbo.careplans	reasoncode	varchar	9	30199	30199	0.050796384
17 dbo.careplans	reasondescription	varchar	56	30199	30199	0.050796384
18						

◀ ▶
Overview
dbo.allergies
dbo.carepla ...
+
⋮
◀ ▶

Figure 6.3: □□□□□□□□□□□□□□□□□□□□□□□□

A	B
1 Sex	Frequency
2	2 61491
3	1 35401
4	List truncated...

Figure 6.4: □□□□□□□□□□□□

Rabbit 1 2 6.4

6.1.2 Rabbit-In-a-Hat

ETL

ETL

ETL

CDM PERSON

Person

Synthea patients 2016.6. patients CDM PERSON

?? Synthea patients CDM PERSON
Field CDM Source
field CDM patients
& comments

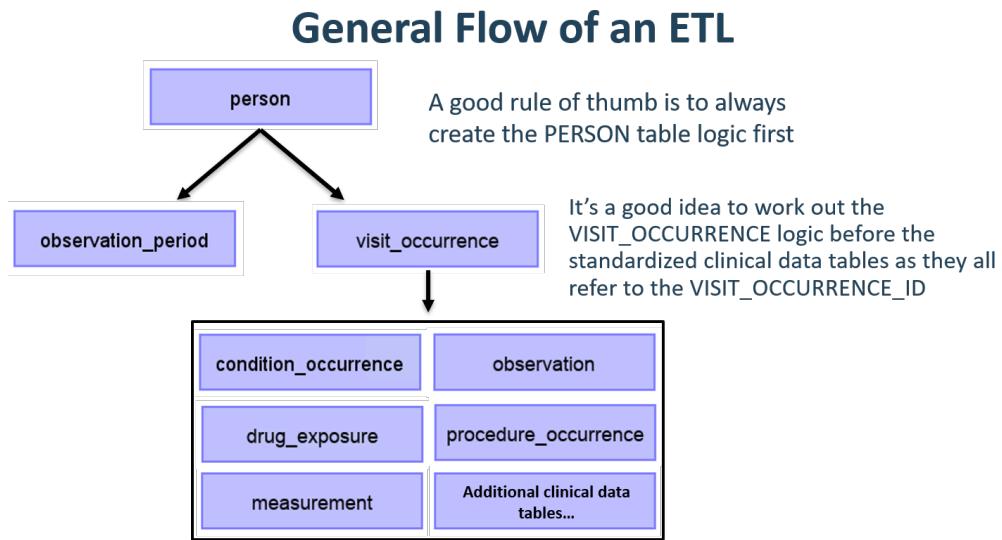


Figure 6.5: ETL

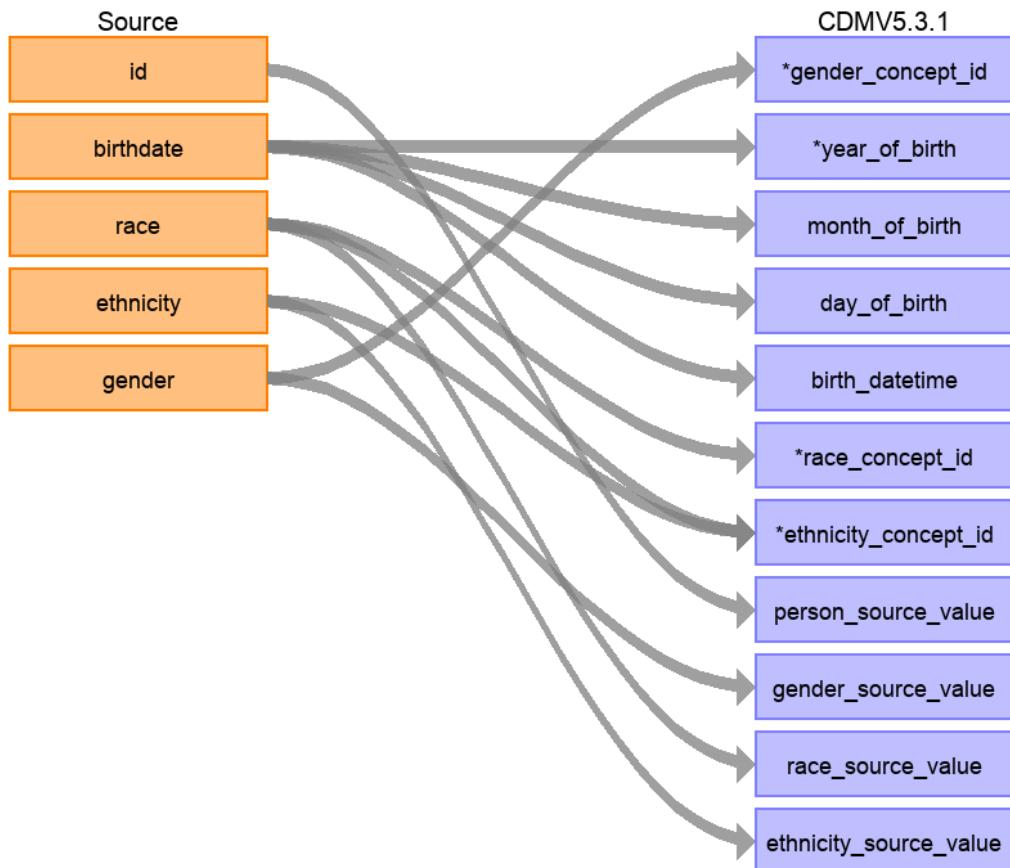


Figure 6.6: Synthea Patients → CDM PERSON

□□(□tab:syntheaEtlPerson) Synthea Patients □□□□□ CDM PERSON □□□□□□□□□ ETL

6.2 2:

```
SELECT concept_id_2 AS standard_concept_id
FROM concept_relationship
INNER JOIN concept AS source_concept
    ON concept_id = concept_id_1
WHERE concept_code = 'I21'
    AND vocabulary_id = 'ICD10CM'
    AND relationship_id = 'Maps to';
```

STANDARD_CONCEPT_ID

6.2.1 Usagi

Usagi
Translate³ Usagi

100

□□□□□□□□□□□□□□□□Usagi□□□□□□□□□□□□□□□□

³<https://translate.google.com/>

⁴<https://github.com/OHDSI/Usagi>

A horizontal row of 20 empty square boxes, likely for drawing or writing practice.

□ □ □ □ □ □ □ **Usagi** □ □ □ □ □ □

□□□□□□CSV□□□Excel (.xlsx)

Translate

File → Import codes... 6.7 Import

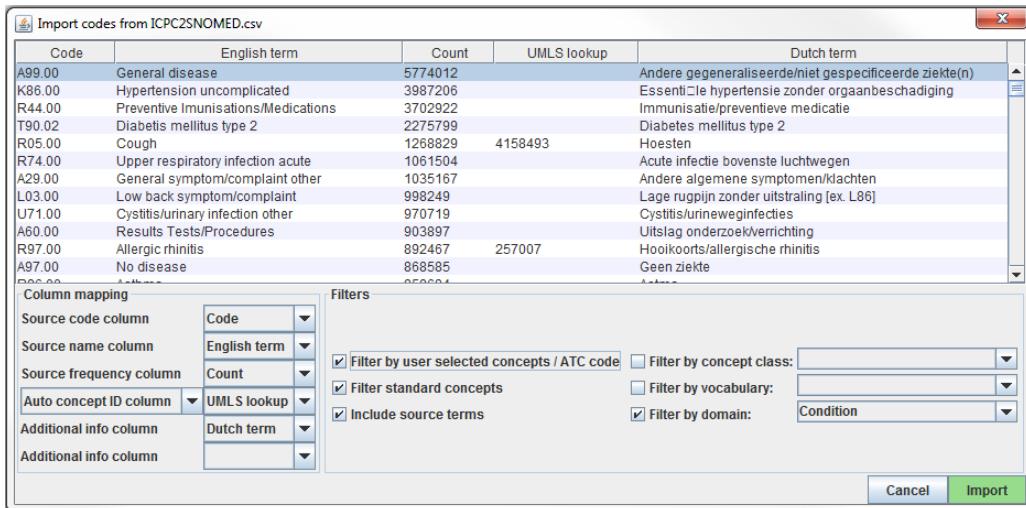


Figure 6.7: Usagi

Column mapping info Usagi info

1. Column mapping “Auto concept ID column” ATC column”
 2. Column mapping ATC AT

6.3 3: ETL

ETL ETL
6.6

ETL

- ETL-Synthea - Synthea
 - <https://github.com/OHDSI/etl-synthea>
 - ETL-CDMBuilder -
 - <https://github.com/OHDSI/etl-cdmbuilder>
 - ETL-LambdaBuilder - AWS lambda
 - <https://github.com/OHDSI/etl-lambdabuilder>

ETL

6.4 4:

>□□□□□□□->□□□□□□□->□□□□□□□ CDM □□□□□□□□□□□□□

- ETL
– CDM
 - CDM
– ETL
 - CDM
– ETL
 - CDM
– ETL
 - CDM
– ETL

ETL OHDSI

15 □□□□□□□□□□

6.5 ETL THEMIS

CDM ETL

OHDSI OHDSI

Wiki CDM ETL

⁵<https://github.com/OHDSI/CommonDataModel/wiki>

Wiki Themis

6.6 CDM ETL

ETL□□□□□□□□□□□□□□□□ ETL□□□□□□□□□□□□□□□□

CRM ETI

6.7 ETL

ETL

- 80/20
 -
 - CDM ETL
 - OHDSI CDM

6.8



- ETL
 - * CDM
 - * OHDSI
 - * ETL
 - * OHDSI

6.9

6.1. ETL

- A) □□□□□□□ CDM □□□□□□□ ETL □□□□
B) □□□ ETL □□□

⁶<https://github.com/OHDSI/Themis>

⁷<http://forums.ohdsi.org/>

⁸<https://forums.ohdsi.org/c/implementers>

- C) □□□□□□□□□□□□□□□□□
D) □□□□□□□□□□□□□□□

6.2. OHDSI PERSON

□: PERSON □□□□□

PERSON_ID	A123B456
GENDER_CONCEPT_ID	8532
YEAR_OF_BIRTH	NULL
MONTH_OF_BIRTH	NULL
DAY_OF_BIRTH	NULL
RACE_CONCEPT_ID	0
ETHNICITY_CONCEPT_ID	8527
PERSON_SOURCE_VALUE	A123B456
GENDER_SOURCE_VALUE	F
RACE_SOURCE_VALUE	WHITE
ETHNICITY_SOURCE_VALUE	□□□□□□□□□□

□□ 6.3. VISIT_OCCURRENCE □□□□□□□□□□□□□□□ Synthea □□□□□□□
PATIENT □ START □ END □□□□□□□□□□□□□ PERSON_ID □□□□□□□ END □□□□

- MIN(START) □ VISIT_START_DATE □ □ □ □
 - MAX(END) □ VISIT_END_DATE □ □ □ □
 - “IP” □ PLACE_OF_SERVICE_SOURCE_VALUE □ □ □ □

6.8 CDM VISIT OCCURRENCE

Data Output						Explain	Messages	Notifications	Query History
	id character varying (1000)	start date	stop date	patient character varying (1000)	encounterclass character varying (1000)				
1	12	2004-09-26	2004-09-27	11	inpatient				
2	13	2004-09-27	2004-09-30	11	inpatient				

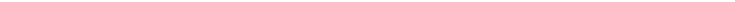
Figure 6.8: □□□□□□□□□

E.3

7 - -

Chapter 7

□□□□□□□□: *David Madigan*

OHDSI 

7.1

A horizontal row of fifteen empty rectangular boxes, intended for children to write their names in, likely as part of a classroom activity.

A horizontal row of 20 small, empty rectangular boxes, likely used for input fields or placeholder text in a form.

A horizontal row of 20 empty square boxes, likely for grading student responses.

- □□□□□□□...□
 - □□□□□□□□□□...□
 - □□□□□□...□
 - □□□□□□□□□□□□□□□□□□□□□□...□
 - □□□HbA1c□□□...□
 - □□□□□□□...□

- □□□□□□□□□□□□□□
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 - □□□□□□□□
 - □□□□□□□□□
 - □□□□□□□□

7.2

A horizontal row of eight empty rectangular boxes, likely used for input fields or placeholder text in a form.

A horizontal row of 20 small, empty rectangular boxes, likely used for a survey or form to indicate presence or selection.

A decorative horizontal bar consisting of a series of small, evenly spaced rectangular blocks, likely made of wood or a similar material, arranged in a straight line.

- ...□□□□□
 - □□...□□□□□□
 - □□□□□□□□□□
 - Y□□□□X□□□□□
 - ...□□□□□□□□□

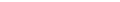
A horizontal row of 15 empty rectangular boxes, likely used for input fields or placeholder text in a form.

- -
 -
 -
 -
 -
 -
 -

7.3

A horizontal row of 20 empty square boxes, intended for students to write their answers in a handwriting practice exercise.

A horizontal row of 20 empty square boxes, likely used for grading student responses.

- 
 - 

- -
 - /
 -



7.4

ACE

7.4.1

10 Chapter 11

7.4.2

Chapter 12 ACE

7.4.3

13 ACE 1

7.5

OHDSI

7.5.1

7.5.2

OHDSI perkins2017principled

76



77

7.1.

□□□□□□□□□ Appendix E.4 □□□□□

Chapter 8

8 - - OHDSI

□□□□□ *Lead: Martijn Schuemie & Frank DeFalco*

OHDSI

8.1

□ 8.1 □ CDM □

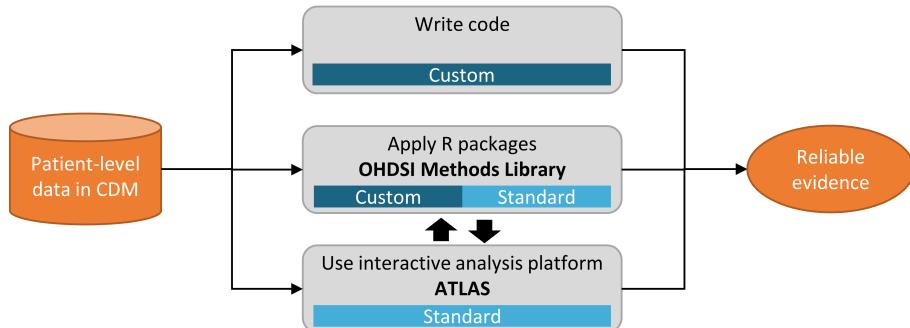
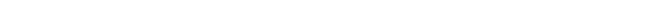
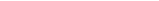


Figure 8.1: CDM

30 ATLAS Libraries

Library  ATLAS 

ATLAS Methods Library

8.2

CDM Methods Library

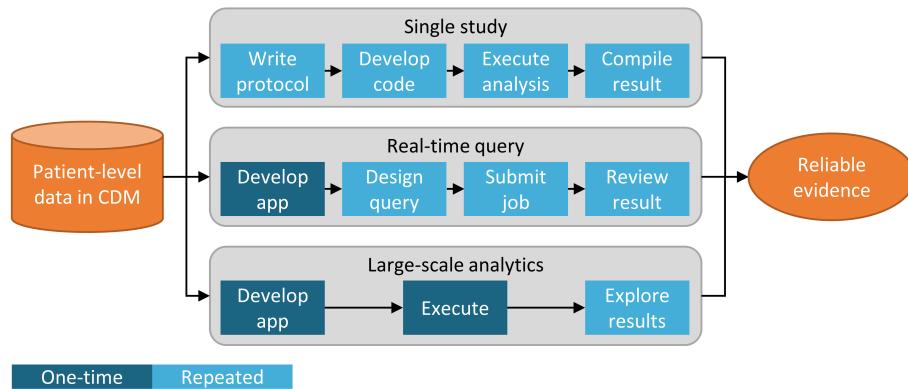


Figure 8.2: (□□□)□□□□□□□□□□□□□□□□□□□□

et al., 2017,) OHDSI Methods Library

et al., 2018b,)□ □ □ □ □ □ □ //data.ohdsi.org/SystematicEvidence/

8.3 ATLAS

ATLAS CDM OHDSI
WebAPI Apache Tom-
cat CDM

8.3 ATLAS ATLAS

ATLAS OMOP

¹<http://www.ohdsi.org/web/atlas>

The screenshot shows the ATLAS software interface for defining a cohort. The left sidebar contains navigation links: Home, Data Sources, Search, Concept Sets, Cohort Definitions (selected), Characterizations, Cohort Pathways, Incidence Rates, Profiles, Estimation, Prediction, Jobs, Configuration, and Feedback. The Cohort Definitions section is active, indicated by a blue background.

The main area displays a "Cohort #1770710" titled "New users of ACE inhibitors as first-line monotherapy for hypertension". Below the title are tabs: Definition (selected), Concept Sets, Generation, Reporting, Export, and Messages (with 3 notifications). To the right are icons for save, cancel, print, export, and refresh.

The "Definition" tab shows a text input field: "enter a cohort definition description here".

The "Cohort Entry Events" section includes a "Help" icon and a "Add Initial Event" button. It lists criteria: "Events having any of the following criteria:" followed by "a drug exposure of ACE inhibitors" with a dropdown arrow, a "+ Add attribute" button, and a "Delete Criteria" button. A note below says "X for the first time in the person's history".

Below this, it says "with continuous observation of at least 365 days before and 0 days after event index date" and "Limit initial events to earliest event per person." A "Restrict initial events" button is present.

The "Inclusion Criteria" section has a "Help" icon and a "New inclusion criteria" button. It lists: 1. has hypertension diagnosis in 1 yr prior to treatment and 2. Has no prior antihypertensive drug exposures in medical.

The bottom left corner features the Apache 2.0 logo and the text "open source software". The bottom right corner shows the OHDSI logo with the tagline "join the journey".

Figure 8.3: ATLAS

8.3.1

ATLAS WebAPI Shiro WebAPI

²<https://github.com/OHDSI/WebAPI/wiki/Security-Configuration>

8.3.2

ATLAS  ATLAS GitHub  wiki        <img alt="GitHub logo" data-bbox="40315 885 403

8.3.3

ATLAS OHDSI WebAPI GitHub⁴ WebAPI GitHub

8.4

OHDSI 8.4 R

8.4.1

Chapter
18

8.4.2

³<https://github.com/OHDSI/ATLAS/wiki>

⁴<https://github.com/OHDSI/Atlas/wiki/Atlas-Setup-Guide>

Prediction and estimation methods	<p>Cohort Method New-user cohort studies using large-scale regression for propensity and outcome models</p>	<p>Self-Controlled Case Series Self-Controlled Case Series analysis using few or many predictors, includes splines for age and seasonality.</p>	<p>Self-Controlled Cohort A self-controlled cohort design, where time preceding exposure is used as control.</p>
	<p>Patient Level Prediction Build and evaluate predictive models for user-specified outcomes, using a wide array of machine learning algorithms.</p>	<p>Case-control Case-control studies, matching controls on age, gender, provider, and visit date. Allows nesting of the study in another cohort.</p>	<p>Case-crossover Case-crossover design including the option to adjust for time-trends in exposures (so-called case-time-control).</p>
Method characterization	<p>Empirical Calibration Use negative control exposure-outcome pairs to profile and calibrate a particular analysis design.</p>	<p>Method Evaluation Use real data and established reference sets as well as simulations injected in real data to evaluate the performance of methods.</p>	<p>Evidence Synthesis Combining study diagnostics and results across multiple sites.</p>
Supporting packages	<p>Database Connector Connect directly to a wide range of database platforms, including SQL Server, Oracle, and PostgreSQL.</p>	<p>Sql Render Generate SQL on the fly for the various SQL dialects.</p>	<p>Cyclops Highly efficient implementation of regularized logistic, Poisson and Cox regression.</p>
	<p>ParallelLogger Support for parallel computation with logging to console, disk, or e-mail.</p>	<p>Feature Extraction Automatically extract large sets of features for user-specified cohorts using data in the CDM.</p>	

Figure 8.4: OHDSI □□□□□□□□□□□□□□□□□

8.4.3

R Library⁵ GitHub CRAN R
Library GitHub CRAN

8.4.4

CDM
SQL Server Oracle Microsoft APS IBM Netezza Amazon
Redshift Impala Hadoop Google Big-Query
Server R RStudio
Library Java

8.4.5

OHDSI R

1. R
2. Rtools Windows R
3. RStudio R IDE

8.5

ATLAS Methods Library OHDSI Web Services (AWS)

8.5.1 Broadsea

Broadsea⁷ Docker [^^dockerUrl] OHDSI Docker
Windows MacOS Linux Broadsea
Docker Methods Library ATLAS OHDSI

⁵<https://ohdsi.github.io/MethodsLibrary>

⁶<https://ohdsi.github.io/CohortMethod/articles/MultipleAnalyses.html>

⁷<https://github.com/OHDSI/Broadsea>

8.5.2 Amazon AWS

Amazon AWS in-a-Box⁸ OHDSI on AWS [^ohdsiOnAwsUrl]

OHDSI-in-a-Box OHDSI CDM OHDSI-in-a-Box OHDSI-in-a-Box ETL OHDSI-in-a-Box OHDSI-in-a-Box 8.5

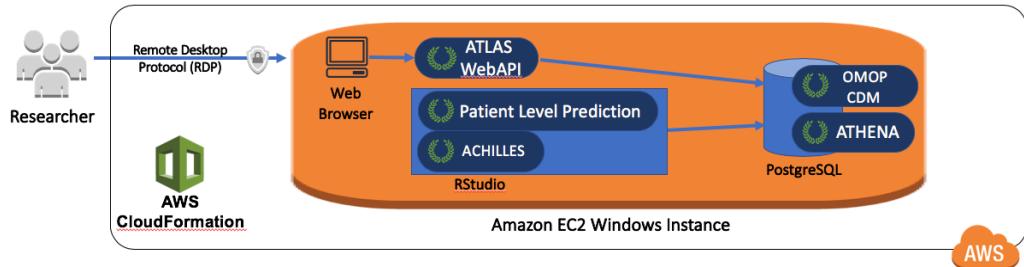


Figure 8.5: OHDSI-in-a-Box □ Amazon Web Services □ □ □ □ □ □ □

OHDSIonAWS
Redshift
Server

RStudio
OHDSI Methods Library

OHDSI
ATLAS

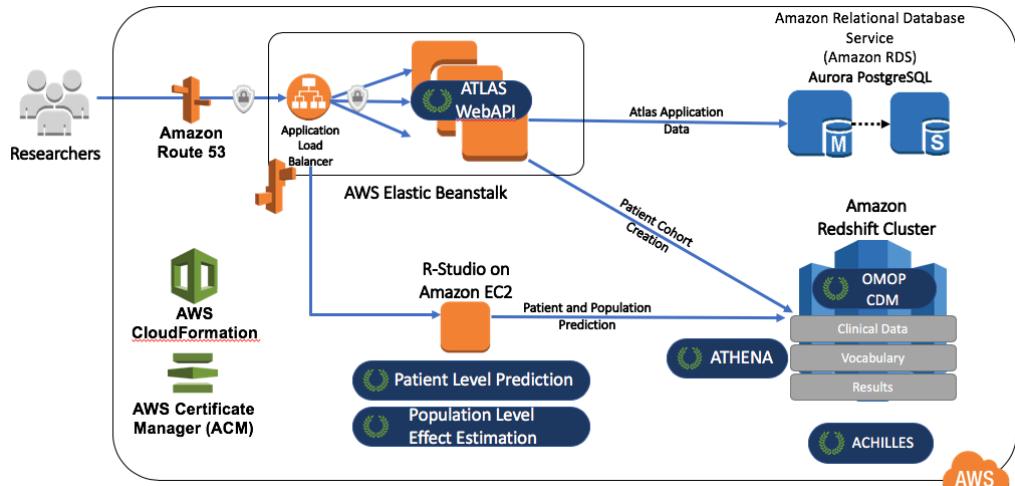


Figure 8.6: OHDSIonAWS □ Amazon Web Services □ □ □ □ □ □ □

⁸<https://github.com/OHDSI/OHDSI-in-a-Box>

8.6



- CDM
 - * CDM
 - * OHDSI Methods Library R
 - * CDM ATLAS
 - OHDSI
 - * CDM
 - * OHDSI
 - * CDM
 - OHDSI OHDSI
 - * CDM ATLAS
 - * OHDSI Methods Library R
 - OHDSI

Chapter 9

9 – – – SQL R

□□□□□□□: Martijn Schuemie & Peter Rijnbeek

CDM
SQL Server
Library
ATI
OHDSI
SQL
Structured
Query
Language
SQL
Server
PERSON
10

```
SELECT TOP 10 * FROM person;
```

PostgreSQL

```
SELECT * FROM person LIMIT 10;
```

OHDSI - OHDSI SQL - SQL Server SQL - SQL OHDSI SQL

SQL

CDM QHDSI

SQL SQLRender Database CDM OHDSI

9.1 SqIRender

SqlRender  CRAN Comprehensive R Archive Network

```
install.packages("SqlRender")
```

9.1.1 SQL

SQL

a render

```
sql <- "SELECT * FROM concept WHERE concept_id = @a;"  
render(sql, a = 123)
```

```
## [1] "SELECT * FROM concept WHERE concept_id = 123;"
```

```
sql <- "SELECT * FROM @x WHERE person_id = @a;"  
render(sql, x = "observation", a = 123)
```

```
## [1] "SELECT * FROM observation WHERE person_id = 123;"
```

```
sql <- "SELECT * FROM concept WHERE concept_id IN (@a);"
render(sql, a = c(123, 234, 345))
```

```
## [1] "SELECT * FROM concept WHERE concept_id IN (123,234,345);"
```

If-Then-Else

```
{Condition} ? {if true} : {if false}    condition  
true 1 iftrue 0 iffase
```

```
sql <- "SELECT * FROM cohort {@x} ? {WHERE subject_id = 1}"
render(sql, x = FALSE)
```

```
## [1] "SELECT * FROM cohort "
```

```
render(sql, x = TRUE)
```

```
## [1] "SELECT * FROM cohort WHERE subject_id = 1"
```

```
□□□□□□□□□□□□□□□□□
```

```
sql <- "SELECT * FROM cohort {@x == 1} ? {WHERE subject_id = 1};"
render(sql, x = 1)
```

```
## [1] "SELECT * FROM cohort WHERE subject_id = 1;"
```

```
render(sql, x = 2)
```

```
## [1] "SELECT * FROM cohort ;"
```

```
IN □□□□□□□□□□□□□□□
```

```
sql <- "SELECT * FROM cohort {@x IN (1,2,3)} ? {WHERE subject_id = 1};"
render(sql, x = 2)
```

```
## [1] "SELECT * FROM cohort WHERE subject_id = 1;"
```

9.1.2 SQL

SqlRender □□□□□□□□□□□□□□□□□ OHDSI SQL □□□□ SQL □□□□□□□□□□□□□□□

```
sql <- "SELECT TOP 10 * FROM person;"
translate(sql, targetDialect = "postgresql")
```

```
## [1] "SELECT * FROM person LIMIT 10;"
## attr("sqlDialect")
## [1] "postgresql"
```

```
targetDialect □□□□□□□□□□□□□□□□□ oracle, postgresql, pdw, redshift, impala, netezza, bigquery, sqlite, sql server
```



□□□□SQL□□□□
SQL□□□□□SQL□□□
Server□□□□□□□□□

SQL OHDSI

Translate

SQL Server

supported by translate

ABS	EXP	RAND
ACOS	FLOOR	RANK
ASIN	GETDATE	RIGHT
ATAN	HASHBYTES*	ROUND
AVG	ISNULL	ROW_NUMBER
CAST	ISNUMERIC	RTRIM
CEILING	LEFT	SIN
CHARINDEX	LEN	SQRT
CONCAT	LOG	SQUARE
COS	LOG10	STDEV
COUNT	LOWER	SUM
COUNT_BIG	LTRIM	TAN
DATEADD	MAX	UPPER
DATEDIFF	MIN	VAR
DATEFROMPARTS	MONTH	YEAR
DATETIMEFROMPARTS	NEWID	
DAY	PI	
EOMONTH	POWER	

* Oracle  SQLite 

SOL

```
-- Simple selects:  
SELECT * FROM table;  
  
-- Selects with joins:  
SELECT * FROM table_1 INNER JOIN table_2 ON a = b;  
  
-- Nested queries:
```

```
SELECT * FROM (SELECT * FROM table_1) tmp WHERE a = b;

-- Limiting to top rows:
SELECT TOP 10 * FROM table;

-- Selecting into a new table:
SELECT * INTO new_table FROM table;

-- Creating tables:
CREATE TABLE table (field INT);

-- Inserting verbatim values:
INSERT INTO other_table (field_1) VALUES (1);

-- Inserting from SELECT:
INSERT INTO other_table (field_1) SELECT value FROM table;

-- Simple drop commands:
DROP TABLE table;

-- Drop table if it exists:
IF OBJECT_ID('ACHILLES_analysis', 'U') IS NOT NULL
    DROP TABLE ACHILLES_analysis;

-- Drop temp table if it exists:
IF OBJECT_ID('tempdb..#cohorts', 'U') IS NOT NULL
    DROP TABLE #cohorts;

-- Common table expressions:
WITH cte AS (SELECT * FROM table) SELECT * FROM cte;

-- OVER clauses:
SELECT ROW_NUMBER() OVER (PARTITION BY a ORDER BY b)
    AS "Row Number" FROM table;

-- CASE WHEN clauses:
SELECT CASE WHEN a=1 THEN a ELSE 0 END AS value FROM table;

-- UNIONs:
SELECT * FROM a UNION SELECT * FROM b;

-- INTERSECTIONS:
SELECT * FROM a INTERSECT SELECT * FROM b;

-- EXCEPT:
SELECT * FROM a EXCEPT SELECT * FROM b;
```

□ □ □ □ □ □

```
SQL Server SELECT first_name + ' ' + last_name AS full_name FROM table  
PostgreSQL Oracle SELECT first_name || ' ' || last_name AS full_name FROM table  
SqlRender SELECT first_name + last_name AS full_name FROM table  
SqlRender last_name + CAST(age AS VARCHAR(3)) AS full_name FROM table
```

□ □ □ □ □ □ □ □ □ AS □ □ □ □

SQL Server PostgreSQL Redshift AS

```
-- Using AS keyword
SELECT *
FROM my_table AS table_1
INNER JOIN (
    SELECT * FROM other_table
) AS table_2
ON table_1.person_id = table_2.person_id

-- Not using AS keyword
SELECT *
FROM my_table table_1
INNER JOIN (
    SELECT * FROM other_table
) table_2
ON table_1.person_id = table_2.person_id
```

Oracle AS SqlRend AS

6 / 6

100

```
sql <- "SELECT * FROM #children;"  
translate(sql, targetDialect = "oracle", oracleTempSchema = "temp_schema")
```

```
## Warning: The 'oracleTempSchema' argument is deprecated. Use 'tempEmulation'
## This warning is displayed once every 8 hours.

## [1] "SELECT * FROM temp_schema.n0hfc14ochildren ;"
## attr(,"sqlDialect")
## [1] "oracle"

temp_schema Oracle 30 Oracle 22 Oracle
Oracle Oracle Oracle
TRUNCATE DROP Oracle

SQL Server Server
```

```
CREATE TABLE #temp (txt VARCHAR);

INSERT INTO #temp
SELECT '1';

SELECT * FROM #temp WHERE txt = 1;
```

```
SELECT * FROM #temp WHERE txt = CAST(1 AS VARCHAR);
```

1

```
SELECT * FROM #temp WHERE CAST(txt AS INT) = 1;
```

A horizontal row of fifteen empty rectangular boxes, intended for children to write their names in, likely as part of a classroom activity.

SOL Server DBMS

```
SELECT * FROM concept WHERE concept_class_id = 'Clinical Finding'
```

A horizontal row of 20 empty square boxes for writing responses.

```
SELECT * FROM concept WHERE LOWER(concept_class_id) = 'clinical finding'
```

SQL Server → cdm_data.dbo.person
→ cdm_data.dbo.person
Server → 1.dbo
PostgreSQL → SQL Server

```
SELECT * FROM @databaseSchema.person
```

```
SQL Server databaseSchema  
= "cdm_data.dbo"  
= "cdm_data"
```

```
USE cdm_data.dbo;
```

SQL

SQL

SQL Shiny
SqlRender

```
launchSqlRenderDeveloper()
```

““`{r sqlDeveloper, fig.cap=‘The SqlDeveloper’}

9.2 DatabaseConnector

```
install.packages("DatabaseConnector")
```

9.2.1

```
conn <- connect(dbms = "postgresql",
                 server = "localhost/postgres",
                 user = "joe",
                 password = "secret",
                 schema = "cdm")
```

```
## Connecting using PostgreSQL driver
```

?connect

`disconnect(conn)`

□□□□□□□□□□□□□□ JDBC □□□□□□□□□□□□□□

```
connString <- "jdbc:postgresql://localhost:5432/postgres"  
conn <- connect(dbms = "postgresql",  
                 connectionString = connString,  
                 user = "joe",  
                 password = "secret",  
                 schema = "cdm")
```

```
## Connecting using PostgreSQL driver
```

```
details <- createConnectionDetails(dbms = "postgresql",
                                    server = "localhost/postgres",
                                    user = "joe",
                                    password = "secret",
                                    schema = "cdm")
conn <- connect(details)
```

```
## Connecting using PostgreSQL driver
```

9.2.2

querySql executeSql querySql

□ □ □ □ □ □

```
querySql(conn, "SELECT TOP 3 * FROM person")
```

```
##   person_id gender_concept_id year_of_birth
## 1          1            8507    1975
## 2          2            8507    1976
## 3          3            8507    1977
```

```
executeSql(conn, "TRUNCATE TABLE foo; DROP TABLE foo;")
```

9.2.3 ffdf

```
x <- querySql.ffdf(conn, "SELECT * FROM person")
```

x□□□ffdf□□□□□□□□□□

9.2.4 SQL

```
SqlRender → render → translate → renderTransl...
```

```
x <- renderTranslateQuerySql(conn,
                           sql = "SELECT TOP 10 * FROM @schema.person",
                           schema = "cdm_synpuf")
```

SQL Server TOP 10 PostgreSQL LIMIT 10 SQL @schema

9.2.5

executeSql

```
data(mtcars)
insertTable(conn, "mtcars", mtcars, createTable = TRUE)
```

mtcars mtcars

9.3 CDM

OHDSI SOL CDM Q

A horizontal row of 30 empty square boxes, intended for students to write their answers in a grid format.

```
SELECT COUNT(*) AS person_count FROM @cdm.person;
```

PERSON_COUNT

26299001

A horizontal row of 20 empty square boxes, intended for students to draw or write in.

```
SELECT AVG(DATEDIFF(DAY,
                      observation_period_start_date,
                      observation_period_end_date) / 365.25) AS num_years
FROM @cdm.observation_period;
```

NUM_YEARS

1.980803

```
SELECT MAX(YEAR(observation_period_end_date) -  
          year_of_birth) AS max_age  
FROM @cdm.person  
INNER JOIN @cdm.observation_period  
  ON person.person_id = observation_period.person_id;
```

MAX_AGE

... AS “ages” ages < .50 * n

```
WITH ages
AS (
    SELECT age,
           ROW_NUMBER() OVER (
               ORDER BY age
           ) order_nr
    FROM (
        SELECT YEAR(observation_period_start_date) - year_of_birth AS age
        FROM @cdm.person
```

```

        INNER JOIN @cdm.observation_period
          ON person.person_id = observation_period.person_id
    ) age_computed
)
SELECT MIN(age) AS min_age,
MIN(CASE
      WHEN order_nr < .25 * n
        THEN 9999
      ELSE age
    END) AS q25_age,
MIN(CASE
      WHEN order_nr < .50 * n
        THEN 9999
      ELSE age
    END) AS median_age,
MIN(CASE
      WHEN order_nr < .75 * n
        THEN 9999
      ELSE age
    END) AS q75_age,
MAX(age) AS max_age
FROM ages
CROSS JOIN (
  SELECT COUNT(*) AS n
  FROM ages
) population_size;

```

MIN_AGE	Q25_AGE	MEDIAN_AGE	Q75_AGE	MAX_AGE
0	6	17	34	90

SQL R R

```

sql <- "SELECT YEAR(observation_period_start_date) -
        year_of_birth AS age
FROM @cdm.person
INNER JOIN @cdm.observation_period
  ON person.person_id = observation_period.person_id;"
age <- renderTranslateQuerySql(conn, sql, cdm = "cdm")
quantile(age[, 1], c(0, 0.25, 0.5, 0.75, 1))

```

```

##   0% 25% 50% 75% 100%
##   0   6   17   34   90

```

CDM

```
SELECT TOP 10 condition_source_value,
       COUNT(*) AS code_count
  FROM @cdm.condition_occurrence
 GROUP BY condition_source_value
 ORDER BY -COUNT(*);
```

CONDITION_SOURCE_VALUE	CODE_COUNT
4019	49094668
25000	36149139
78099	28908399
319	25798284
31401	22547122
317	22453999
311	19626574
496	19570098
I10	19453451
3180	18973883

□□□ CONDITION OCCURRENCE □□□□□ CONDITION SOURCE VALUE □□□□

9.4

```
SELECT COUNT(*) AS subject_count,
       concept_name
  FROM @cdm.person
 INNER JOIN @cdm.concept
    ON person.gender_concept_id = concept.concept_id
 GROUP BY concept.name;
```

SUBJECT_COUNT	CONCEPT_NAME
14927548	FEMALE
11371453	MALE

```
SELECT COUNT(*) AS prescription_count  
FROM @cdm.drug_exposure  
INNER JOIN @cdm.concept_ancestor
```

```
    ON drug_concept_id = descendant_concept_id
INNER JOIN @cdm.concept ingredient
    ON ancestor_concept_id = ingredient.concept_id
WHERE LOWER(ingredient.concept_name) = 'ibuprofen'
    AND ingredient.concept_class_id = 'Ingredient'
    AND ingredient.standard_concept = 'S';
```

PRESCRIPTION_COUNT

9.5 QueryLibrary

QueryLibrary □ CDM □ SQL □ 1

Select a query	
Column visibility	
Show	10 ▾ entries
Search:	<input type="text"/>
Group	Name
[“drug exposure”]	All
drug exposure	DEX01 Counts of persons with any number of exposures to a certain drug
drug exposure	DEX02 Counts of persons taking a drug, by age, gender, and year of exposure
drug exposure	DEX03 Distribution of age, stratified by drug
drug exposure	DEX04 Distribution of gender in persons taking a drug
drug exposure	DEX05 Counts of drug records for a particular drug
drug exposure	DEX06 Counts of distinct drugs in the database
drug exposure	DEX07 Maximum number of drug exposure events per person over some time period

Query Description

DEX01: Counts of persons with any number of exposures to a certain drug

Description

This query is used to count the persons with at least one exposures to a certain drug (drug_concept_id). See vocabulary queries for obtaining valid drug_concept_id values. The input to the query is a value (or a comma-separated list of values) of a drug_concept_id. If the input is omitted, all drugs in the data table are summarized.

Query

The following is a sample run of the query. The input parameters are highlighted in blue.

```
SELECT
    c.concept_name,
    drug_concept_id,
    COUNT(person_id) AS num_persons
FROM cdm.drug_exposure
INNER JOIN cdm.concept c
ON drug_concept_id = c.concept_id
WHERE drug_concept_id = 'Pain'
```

CDM

¹<http://data.ohdsi.org/QueryLibrary>

²<https://github.com/OHDSI/QueryLibrary>

9.6

9.6.1

9.6.2

9.6.3

9.6.4

9.7 SQL R

OHDSI OHDSI

```
library(DatabaseConnector)
conn <- connect(dbms = "postgresql",
                 server = "localhost/postgres",
                 user = "joe",
                 password = "secret")
cdmDbSchema <- "cdm"
cohortDbSchema <- "scratch"
cohortTable <- "my_cohorts"

sql <- "
CREATE TABLE @cohort_db_schema.@cohort_table (
    cohort_definition_id INT,
    cohort_start_date DATE,
    cohort_end_date DATE,
    subject_id BIGINT
);
"
renderTranslateExecuteSql(conn, sql,
                         cohort_db_schema = cohortDbSchema,
                         cohort_table = cohortTable)
```

9.7.1

COHORT

```

sql <- "
INSERT INTO @cohort_db_schema.@cohort_table (
    cohort_definition_id,
    cohort_start_date,
    cohort_end_date,
    subject_id
)
SELECT 1 AS cohort_definition_id,
    cohort_start_date,
    cohort_end_date,
    subject_id
FROM (
    SELECT drug_era_start_date AS cohort_start_date,
        drug_era_end_date AS cohort_end_date,
        person_id AS subject_id
    FROM (
        SELECT drug_era_start_date,
            drug_era_end_date,
            person_id,
            ROW_NUMBER() OVER (
                PARTITION BY person_id
                ORDER BY drug_era_start_date
            ) order_nr
        FROM @cdm_db_schema.drug_era
        WHERE drug_concept_id = 1308216 --
    ) ordered_exposures
    WHERE order_nr = 1
) first_era
INNER JOIN @cdm_db_schema.observation_period
    ON subject_id = person_id
    AND observation_period_start_date < cohort_start_date
    AND observation_period_end_date > cohort_start_date
WHERE DATEDIFF(DAY,
    observation_period_start_date,
    cohort_start_date) >= 365;
"
renderTranslateExecuteSql(conn, sql,
    cohort_db_schema = cohortDbSchema,
    cohort_table = cohortTable,
    cdm_db_schema = cdmDbSchema)

```

CPM DRUG ERA DRUG EXPOSURE

9.7.2

A horizontal row of 20 empty square boxes, intended for children to write their names in, likely as part of a classroom activity or name recognition exercise.

□□□ CONDITION OCCURRENCE □□□ CONCEPT ANCESTOR □□□□□

9.7.3

A horizontal row of 20 small, empty rectangular boxes, likely used for input fields or placeholder text in a form.

```
sql <- "
WITH tar AS (
    SELECT concept_name AS gender,
           FLOOR((YEAR(cohort_start_date) -
                  year_of_birth) / 10) AS age,
```

```

subject_id,
cohort_start_date,
CASE WHEN DATEADD(DAY, 7, cohort_start_date) >
    observation_period_end_date
THEN observation_period_end_date
ELSE DATEADD(DAY, 7, cohort_start_date)
END AS cohort_end_date
FROM @cohort_db_schema.@cohort_table
INNER JOIN @cdm_db_schema.observation_period
    ON subject_id = observation_period.person_id
        AND observation_period_start_date < cohort_start_date
        AND observation_period_end_date > cohort_start_date
INNER JOIN @cdm_db_schema.person
    ON subject_id = person.person_id
INNER JOIN @cdm_db_schema.concept
    ON gender_concept_id = concept_id
WHERE cohort_definition_id = 1 --
)
SELECT days.gender,
    days.age,
    days,
    CASE WHEN events IS NULL THEN 0 ELSE events END AS events
FROM (
    SELECT gender,
        age,
        SUM(DATEDIFF(DAY, cohort_start_date,
            cohort_end_date)) AS days
    FROM tar
    GROUP BY gender,
        age
) days
LEFT JOIN (
    SELECT gender,
        age,
        COUNT(*) AS events
    FROM tar
    INNER JOIN @cohort_db_schema.@cohort_table angioedema
        ON tar.subject_id = angioedema.subject_id
            AND tar.cohort_start_date <= angioedema.cohort_start_date
            AND tar.cohort_end_date >= angioedema.cohort_start_date
    WHERE cohort_definition_id = 2 --
    GROUP BY gender,
        age
) events
ON days.gender = events.gender
    AND days.age = events.age;
"
results <- renderTranslateQuerySql(conn, sql,

```

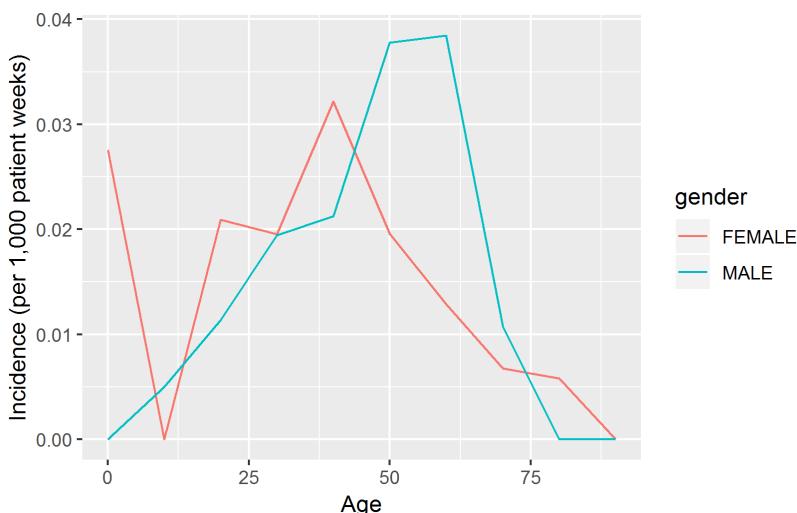
```
cohort_db_schema = cohortDbSchema,
cohort_table = cohortTable,
cdm_db_schema = cdmDbSchema,
snakeCaseToCamelCase = TRUE)
```

□ CTE tar □ SQL □ SQL □ R □
 ggplot2 □

```
#   IR
results$ir <- 1000 * results$events / results$days / 7

#
results$age <- results$age * 10

library(ggplot2)
ggplot(results, aes(x = age, y = ir, group = gender, color = gender)) +
  geom_line() +
  xlab(" ") +
  ylab(" 1,000      ")
```



9.7.4

□ TRUNCATE TABLE @cohort_db_schema.@cohort_table;
 DROP TABLE @cohort_db_schema.@cohort_table;

```
sql <- "
TRUNCATE TABLE @cohort_db_schema.@cohort_table;
DROP TABLE @cohort_db_schema.@cohort_table;"
```

```
""
renderTranslateExecuteSql(conn, sql,
                          cohort_db_schema = cohortDbSchema,
                          cohort_table = cohortTable)

disconnect(conn)
```

9.7.5

OHDSI SQL DatabaseConnector SQLRender SQL SQLSQLRender DatabaseConnector

9.8



- SQL Structured Query Language CDM
- SQL SQL
- SqlRender DatabaseConnector R CDM
- R SQL OHDSI
- QueryLibrary CDM SQL

9.9

8.4.5

R R-Studio Java
SqlRender DatabaseConnector Eunomia

```
install.packages(c("SqlRender", "DatabaseConnector", "remotes"))
remotes::install_github("ohdsi/Eunomia", ref = "v1.0.0")
```

Eunomia CDM R

```
connectionDetails <- Eunomia::getEunomiaConnectionDetails()
```

CDM main

9.1. SQL R

□□□□□□□□□ E.5 □□□□□□

Chapter 10

10 — —

□□□□□: *Kristin Kostka!*

ATLAS SQL

10.1

OHDSI



OHDSI
9/ICD-10
9/ICD-10

A horizontal row of 20 small, empty rectangular boxes, likely used for a grid or table structure.

10.2

- CDM
 -
 - DRUG_EXPOSURE DAYS_SUPPLY
 - 365



Figure 10.1: □□□□□□□□□□



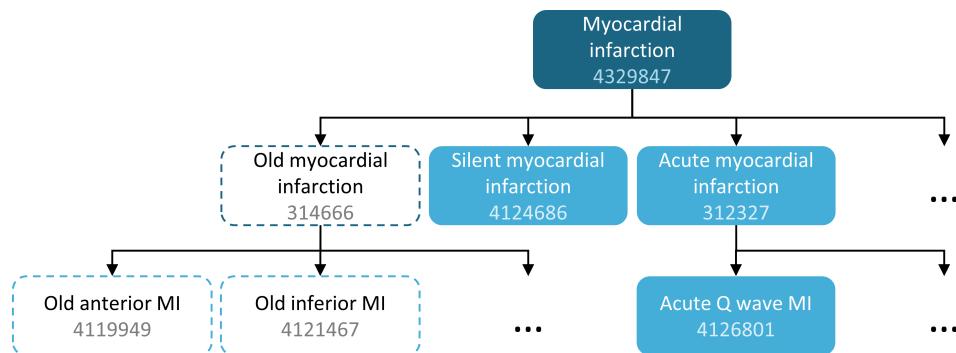
10.3

□□□#tab:conceptSetExpression□□□□□□□□□□□□□□□□□□

□□□□□ID □□□□□□□ □□ □□ □□□□□□□
4329847 □□□□ □□□ □□ □□□
314666 □□□□□□□ □□ □□ □□□

□□□□□□□□□ 10.2 □□□□□

9 □□ ICD-10 □□□□□□□□□□



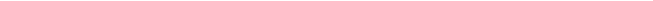
10.4

10.5

¹<https://github.com/OHDSI/Aphrodite>

10.5.1 OHDSI

10.6

10.3 

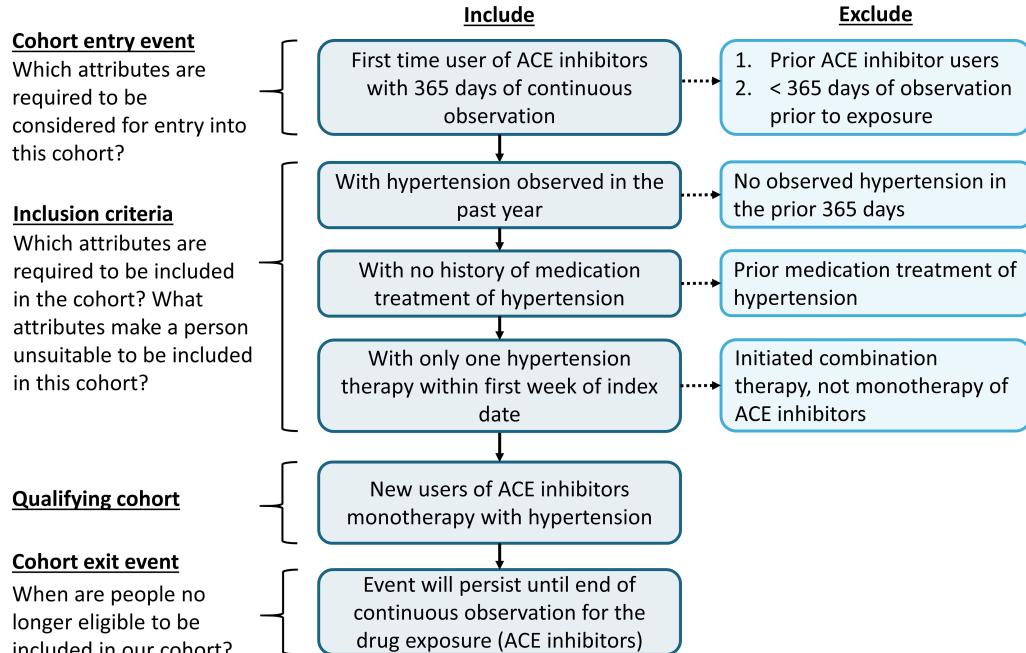


Figure 10.3: □□□□□□□□□□□□□□

 ATLAS CDM

²<https://www.ohdsi.org/web/wiki/doku.php?id=projects:workgroups:gold-library-wg>

10.7 ATLAS

The screenshot shows the ATLAS Cohort Definitions interface. At the top, there is a navigation bar with icons for Home, Cohort Definitions, Cohort Explorer, Reports, and Help. Below the navigation bar, the main title is "Cohort Definitions". The sub-section title is "cohort". A version number "10.4" is displayed on the right side.

The main area is titled "New Cohort Definition". It has tabs for "Definition" (selected), "Concept Sets", "Generation", "Reporting", and "Export". Below the tabs, there is a text input field with placeholder text "enter a cohort definition description here".

The first section is "Cohort Entry Events". It contains a sub-section "Events having any of the following criteria:" with a "Add Initial Event" button. It also includes settings for "with continuous observation of at least [0] days before and [0] days after event index date" and "Limit initial events to: earliest event per person". A "Restrict initial events" button is present.

The second section is "Inclusion Criteria". It has a "New inclusion criteria" button. It includes a note "Limit qualifying events to: earliest event per person".

Figure 10.4: ATLAS

ATLAS New Cohort Definition



ATLAS

ATLAS

10.7.1

Add initial event

10.5 ATLAS CONDITION_OCCURRENCE
Drug Exposure

10.6 ATLAS

10.7.2

ACE

Cohort #1771427

EXAMPLE: new users of ACE inhibitors as first-line mono-therapy for hypertension

Definition Concept Sets Generation Reporting Export

enter a cohort definition description here

Cohort Entry Events

Events having any of the following criteria:

with continuous observation of at least days before and days after event index date.

Limit initial events to: earliest event per person.

+ Add Initial Event

Restrict initial events

Inclusion Criteria

New inclusion criteria

Limit qualifying events to: earliest event per person.

Cohort Exit

Add Condition Era
Find patients with specific diagnosis era.

Add Condition Occurrence
Find patients with specific diagnoses.

Add Death
Find patients based on death.

Add Device Exposure
Find patients based on device exposure.

Add Dose Era
Find patients with dose eras.

Add Drug Era
Find patients with exposure to drugs over time.

Add Drug Exposure
Find patients with exposure to specific drugs or drug classes.

Figure 10.5: □□□□□□□□□

Cohort Entry Events

Events having any of the following criteria:

a drug exposure of **Any Drug**

+ Add attribute... **Delete Criteria**

with continuous observation of at least days before and days after event index date.

Limit initial events to: earliest event per person.

Restrict initial events

Figure 10.6: □□□□□□□

□ □ □ □ 1: □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □

set New Concept Set Un
Concept Set Search ACE

Figure 10.7: □□□□□ - ACE□□□

ATLAS Import Concept Set
inhibitors Any
Drug

10.7.3

attribute ACE Add first exposure criteria



event 10.10

Concept Set Expression Included Concepts 21536 Included Source Codes Export Import

Name: ACE Inhibitors

Show 25 ▾ entries Search: □

Showing 1 to 15 of 15 entries Previous 1 Next

	Concept Id	Concept Code	Concept Name	Domain	Standard Concept Caption	Exclude	Descendants	Mapped
1335471	18867	benazepril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	
1340128	1998	Captopril	Drug	Standard	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	
19050216	21102	Cilazapril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	
1341927	3827	Enalapril	Drug	Standard	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	
1342001	3829	Enalaprilat	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	
1363749	50166	Fosinopril	Drug	Standard	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	
19122327	60245	imidapril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	
1308216	29046	Lisinopril	Drug	Standard	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	
1310756	30131	moexipril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	
1373225	54552	Perindopril	Drug	Standard	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	
1331235	35208	quinapril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	
1334456	35296	Ramipril	Drug	Standard	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	
19040051	36908	spirapril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	
1342439	38454	trandolapril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	
19102107	39990	zofenopril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	

Classification Non-Standard Standard

Figure 10.8: ACE

Import Concept Set From Repository...

New Concept Set

Show 10 ▾ entries Filter Repository Concept Sets: ace inhibitors

ID	Title	Created	Modified	Author
1794480	[OHDSI EU 2019] Excluded concepts of ACE inhibitors or Thiazide diuretics	03/28/2019 11:04 AM	03/28/2019 11:04 AM	anonymous
963	ACE Inhibitors			anonymous
3268	COPY OF: ACE Inhibitors			anonymous
99283	Ace Inhibitors			anonymous
142965	PheKB ACE-I ACE inhibitors			anonymous

Showing 1 to 5 of 5 entries (filtered from 11,667 total entries) Previous 1 Next

Figure 10.9: ATLAS

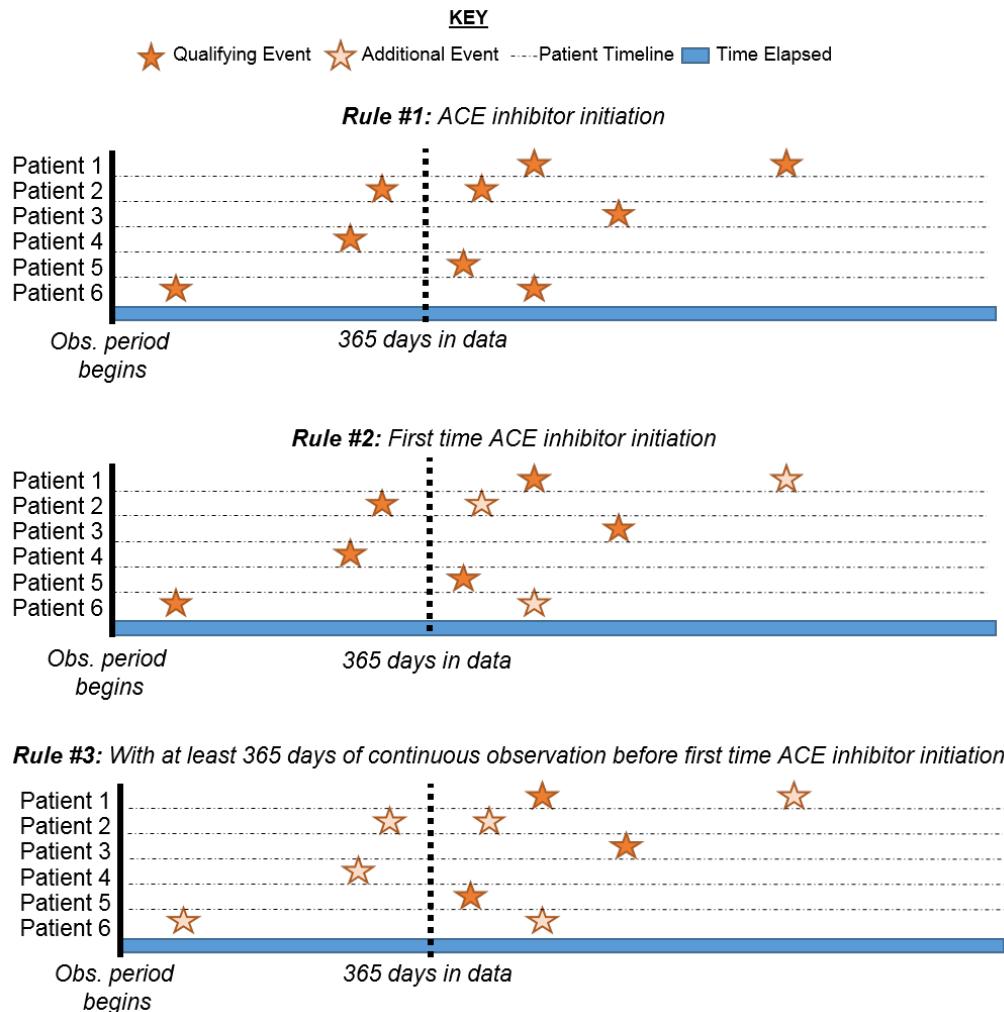


Figure 10.11: □□□□□□□□□□□□□□□

```

password = "supersecret")

cdmDbSchema <- "my_cdm_data"
cohortDbSchema <- "scratch"
cohortTable <- "my_cohorts"
```

3 cdmDbSchema cohortDbSchema cohortTable
SQL Server
<- "my_cdm_data.dbo"

10.8.2

ID R SQL

```

aceI <- c(1308216, 1310756, 1331235, 1334456, 1335471, 1340128, 1341927,
       1342439, 1363749, 1373225)

hypertension <- 316866

allHtDrugs <- c(904542, 907013, 932745, 942350, 956874, 970250, 974166,
                 978555, 991382, 1305447, 1307046, 1307863, 1308216,
                 1308842, 1309068, 1309799, 1310756, 1313200, 1314002,
                 1314577, 1317640, 1317967, 1318137, 1318853, 1319880,
                 1319998, 1322081, 1326012, 1327978, 1328165, 1331235,
                 1332418, 1334456, 1335471, 1338005, 1340128, 1341238,
                 1341927, 1342439, 1344965, 1345858, 1346686, 1346823,
                 1347384, 1350489, 1351557, 1353766, 1353776, 1363053,
                 1363749, 1367500, 1373225, 1373928, 1386957, 1395058,
                 1398937, 40226742, 40235485)
```

10.8.3

ACE

```

conn <- connect(connDetails)

sql <- "SELECT person_id AS subject_id,
      MIN(drug_exposure_start_date) AS cohort_start_date
INTO #first_use
FROM @cdm_db_schema.drug_exposure
INNER JOIN @cdm_db_schema.concept_ancestor
  ON descendant_concept_id = drug_concept_id
WHERE ancestor_concept_id IN (@ace_i)
GROUP BY person_id;"

renderTranslateExecuteSql(conn,
```

```
sql,  
cdm_db_schema = cdmDbSchema,  
ace i = aceI)
```

DRUG EXPOSURE□□□□CONCEPT ANCESTOR□□□□□□□□□□ACE□□□□

10.8.4 365

OBSERVATION PERIOD

```
sql <- "SELECT subject_id,
  cohort_start_date
INTO #has_prior_obs
FROM #first_use
INNER JOIN @cdm_db_schema.observation_period
  ON subject_id = person_id
    AND observation_period_start_date <= cohort_start_date
    AND observation_period_end_date >= cohort_start_date
WHERE DATEADD(DAY, 365, observation_period_start_date) < cohort_start_date;"
```



```
renderTranslateExecuteSql(conn, sql, cdm db schema = cdmDbSchema)
```

10.8.5

365

```
sql <- "SELECT DISTINCT subject_id,
  cohort_start_date
INTO #has_ht
FROM #has_prior_obs
INNER JOIN @cdm_db_schema.condition_occurrence
  ON subject_id = person_id
    AND condition_start_date <= cohort_start_date
    AND condition_start_date >= DATEADD(DAY, -365, cohort_start_date)
INNER JOIN @cdm_db_schema.concept_ancestor
  ON descendant_concept_id = condition_concept_id
WHERE ancestor_concept_id = @hypertension;"
```



```
renderTranslateExecuteSql(conn,
  sql,
  cdm_db_schema = cdmDbsSchema,
  hypertension = hypertension)
```



```
all_ht_drugs = allHtDrugs)
```

10.8.8

```

sql <- "
SELECT person_id,
       CAST(1 AS INT) AS concept_id,
       drug_exposure_start_date AS exposure_start_date,
       drug_exposure_end_date AS exposure_end_date
  INTO #exposure
     FROM @cdm_db_schema.drug_exposure
    INNER JOIN @cdm_db_schema.concept_ancestor
        ON descendant_concept_id = drug_concept_id
   WHERE ancestor_concept_id IN (@ace_i);"
renderTranslateExecuteSql(conn,
                        sql,
                        cdm_db_schema = cdmDbSchema,
                        ace_i = aceI)

```

```
sql <- "
SELECT ends.person_id AS subject_id,
       ends.concept_id AS cohort_definition_id,
       MIN(exposure_start_date) AS cohort_start_date,
       ends.era_end_date AS cohort_end_date
INTO #exposure_era
FROM (
    SELECT exposure.person_id,
           exposure.concept_id,
           exposure.exposure_start_date,
           MIN(events.end_date) AS era_end_date
      FROM #exposure exposure
     JOIN (
--cteEndDates
        SELECT person_id,
               concept_id,
               DATEADD(DAY, - 1 * @max_gap, event_date) AS end_date
      FROM (
          SELECT person_id,
                 concept_id,
                 event_date,
                 event_type,
                 MAX(start ordinal) OVER (

```

```

        PARTITION BY person_id ,concept_id ORDER BY event_date,
        event_type ROWS UNBOUNDED PRECEDING
    ) AS start_ordinal,
ROW_NUMBER() OVER (
    PARTITION BY person_id, concept_id ORDER BY event_date,
    event_type
) AS overall_ord
FROM (
-- select the start dates, assigning a row number to each
    SELECT person_id,
        concept_id,
        exposure_start_date AS event_date,
        0 AS event_type,
        ROW_NUMBER() OVER (
            PARTITION BY person_id, concept_id ORDER BY exposure_start_date
        ) AS start_ordinal
    FROM #exposure exposure

    UNION ALL

-- add the end dates with NULL as the row number, padding the end dates by
-- @max_gap to allow a grace period for overlapping ranges.

    SELECT person_id,
        concept_id,
        DATEADD(day, @max_gap, exposure_end_date),
        1 AS event_type,
        NULL
    FROM #exposure exposure
    ) rawdata
) events
WHERE 2 * events.start_ordinal - events.overall_ord = 0
) events
ON exposure.person_id = events.person_id
    AND exposure.concept_id = events.concept_id
    AND events.end_date >= exposure.exposure_end_date
GROUP BY exposure.person_id,
    exposure.concept_id,
    exposure.exposure_start_date
) ends
GROUP BY ends.person_id,
    concept_id,
    ends.era_end_date;"
```

ACE

```
sql <- "SELECT ee.subject_id,
  CAST(1 AS INT) AS cohort_definition_id,
  ee.cohort_start_date,
  ee.cohort_end_date
INTO @cohort_db_schema.@cohort_table
FROM #monotherapy mt
INNER JOIN #exposure_era ee
  ON mt.subject_id = ee.subject_id
  AND mt.cohort_start_date = ee.cohort_start_date;"

renderTranslateExecuteSql(conn,
                        sql,
                        cohort_db_schema = cohortDbSchema,
                        cohort_table = cohortTable)
```

10.8.9

```
sql <- "TRUNCATE TABLE #first_use;  
DROP TABLE #first_use;  
  
TRUNCATE TABLE #has_prior_obs;  
DROP TABLE #has_prior_obs;  
  
TRUNCATE TABLE #has_ht;  
DROP TABLE #has_ht;  
  
TRUNCATE TABLE #no_prior_ht_drugs;  
DROP TABLE #no_prior_ht_drugs;  
  
TRUNCATE TABLE #monotherapy;  
DROP TABLE #monotherapy;  
  
TRUNCATE TABLE #exposure;  
DROP TABLE #exposure;  
  
TRUNCATE TABLE #exposure_era;  
DROP TABLE #exposure_era;"  
  
renderTranslateExecuteSql(conn, sc  
  
disconnect(conn)
```

10.9



- □□□□□□□□1□□□□□□□□□□□□□□□□□□□□□□□□□□□
 - □□
 - □□**OHDSI**□□□□□□□□□□□□□
 - □□□□□□□□□□□□2□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□
 - □□□□□□□□□□□□□□**ATLAS**□□□**SQL**□□□□□□□□□□□□□

10,10



ATLAS http://atlas-demo.ohdsi.org

□ □ 10.1. □ □ □ □ □ □ □ □ ATLAS □ □ □ □ □ □ □ □ □ □



R-Studio Java
8.4.5 SqlRender DatabaseConnector Eunomia

```
install.packages(c("SqlRender", "DatabaseConnector", "remotes"))
remotes::install_github("ohdsi/Eunomia", ref = "v1.0.0")
```

Eunomia CRM R

```
connectionDetails <- Eunomia:::getEunomiaConnectionDetails()
```

CDM main

□ □ □ 19.2. SQL □ □ □ R □ □ □ COHORT □ □ □ AMI □ □ □

- □□□□□□□□□□□□□□□□4329847□□□□□□□□□□□□□□□□314666

□□□□□□□□□ Appendix E.6 □□□□□□□□□

Chapter 11

11 - -

□□□□: *Anthony Sena & Daniel Prieto-Alhambra*

11.1

A horizontal row of 20 small white squares arranged in a single line.

11.2

OHDSI

the Reporting of Observation Studies in Epidemiology (STROBE) □ STROBE

Elm et al., 2008) □

11.3

et al. (2016) OHDSI 2

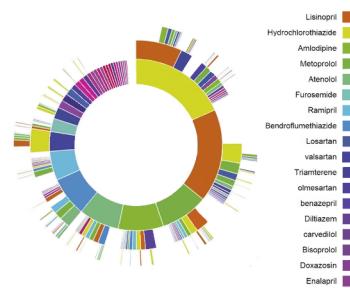
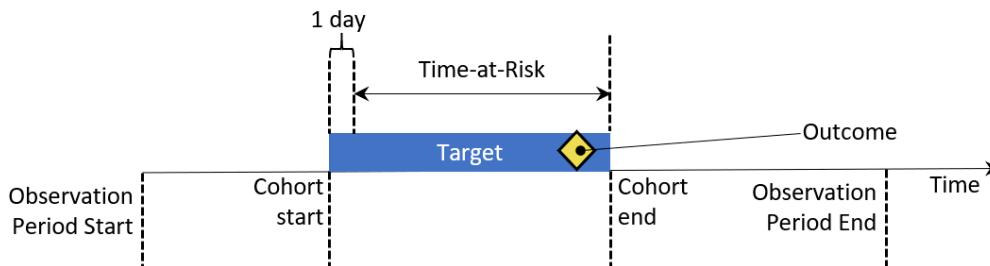


Figure 11.1: OHDSI

11.4

11.2



= #

= #

PUS

11.5

WHO (Who, 2013)

WHO

11.6 ATLAS

Data Sources ATLAS

Occurrence

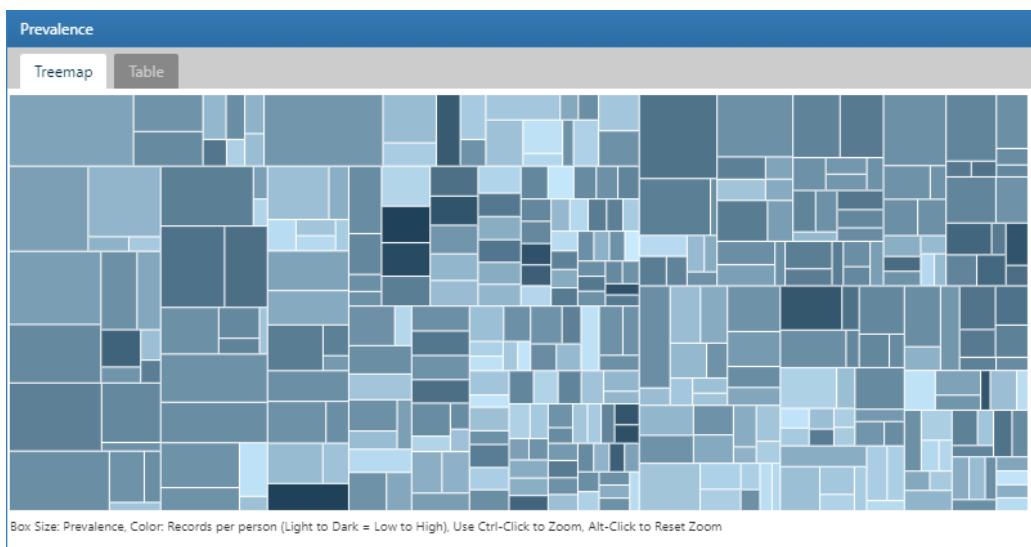


Figure 11.3: Atlas Data Sources: Condition Occurrence Treemap

Prevalence				
Treemap		Table		
		Column visibility	Copy	CSV
		Show 15 ▾	entries	
Showing 1 to 15 of 47 entries (filtered from 15,907 total entries)				
		Filter: hypertension		
		Previous	1 2 3 4	Next
Concept	Name	Person Count	Prevalence	Records per person
Id				
320128	Essential hypertension	17,814,076	12.30%	5.80
312648	Benign essential hypertension	11,014,877	7.61%	4.35
317898	Malignant essential hypertension	1,021,441	0.70%	2.22
381290	Ocular hypertension	521,264	0.36%	2.40
441922	Transient hypertension of pregnancy	209,317	0.14%	2.45
44782429	Chronic kidney disease due to hypertension	170,534	0.12%	3.60
137940	Transient hypertension of pregnancy - delivered	153,806	0.11%	1.07
321080	Hypertension complicating pregnancy, childbirth and the puerperium	148,728	0.10%	2.15
314423	Benign essential hypertension complicating pregnancy, childbirth and the puerperium - not delivered	132,245	0.09%	3.94
44782690	Chronic kidney disease stage 5 due to hypertension	119,375	0.08%	5.20
44783618	Heritable pulmonary arterial hypertension	104,737	0.07%	3.61
319826	Secondary hypertension	96,356	0.07%	2.14
4167493	Pregnancy-induced hypertension	91,675	0.06%	2.60
321074	Pre-existing hypertension complicating pregnancy, childbirth and puerperium	74,311	0.05%	2.99
192680	Portal hypertension	71,240	0.05%	3.11

Figure 11.4: ATLAS

essential hypertension

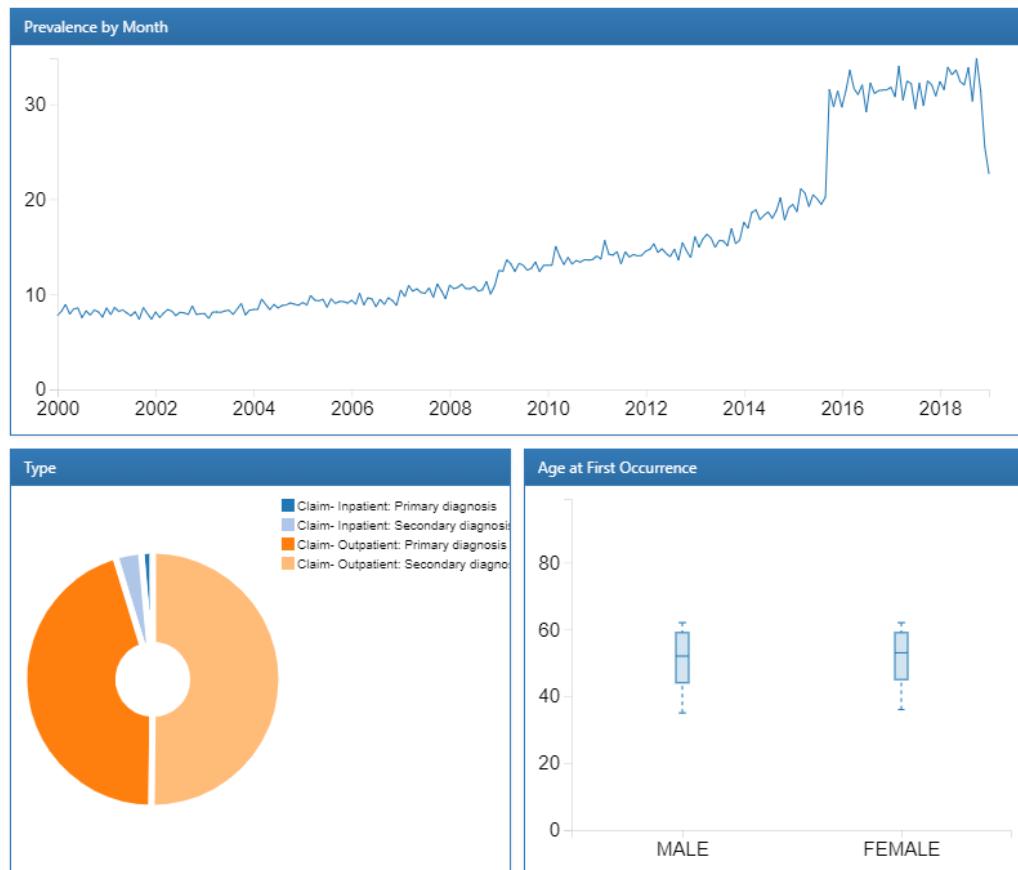


Figure 11.5: ATLAS: Essential Hypertension

Era

11.7 ATLAS

ATLAS

11.7.1

1 1 2 2 3 3
B.6 2 1 3
B.7



Cohort characterization is defined as the process of generating cohort level descriptive summary statistics from person level covariate data. Summary statistics of these person level covariates may be count, mean, sd, var, min, max, median, range, and quantiles. In addition, covariates during a period may be stratified into temporal units of time for time-series analysis such as fixed intervals of time relative to cohort_start_date (e.g. every 7 days, every 30 days etc.), or in absolute calendar intervals such as calendar-week, calendar-month, calendar-quarter, calendar-year.

Cohort definitions

Import

ID	Name		
10447	Patients initiating first-line therapy for hypertension with >1 yr follow-up	Edit cohort	Remove
10448	Patients initiating first-line therapy for hypertension with >3 yr follow-up	Edit cohort	Remove

Show 10 entries Search: Previous 1 Next

Showing 1 to 2 of 2 entries

Figure 11.6: Cohort Characterization - Cohort Definitions

ATLAS

ATLAS Chapter 10 Import 11.6

OMOP CDM

ATLAS OMOP CDM 100 R Feature Extraction R

Import

Demographics Procedure, Condition, Drug

- **Any time prior:** 365
- **Long term:** 365
- **Medium term:** 180
- **Short term:** 30

ATLAS

ATLAS

ATLAS

Feature analyses

Import

Show 25 ▾ entries Search: □

ID	Name	Description	Actions
43	Drug Era Short Term	One covariate per drug in the drug_era table overlapping with any part of the short window.	Remove
49	Charlson Index	The Charlson comorbidity index (Romano adaptation) using all conditions prior to the window end.	Remove
67	Condition Occurrence Long Term	One covariate per condition in the condition_occurrence table starting in the long term window.	Remove
71	Demographics Age Group	Age of the subject on the index date (in 5 year age groups)	Remove
72	Demographics Race	Race of the subject.	Remove
73	Demographics Prior Observation Time	Number of continuous days of observation time preceding the index date.	Remove
74	Demographics Gender	Gender of the subject.	Remove
76	Condition Occurrence Medium Term	One covariate per condition in the condition_occurrence table starting in the medium term window.	Remove
77	Demographics Age	Age of the subject on the index date (in years).	Remove
79	Demographics Time In Cohort	Number of days of observation time during cohort period.	Remove
80	Demographics Index Year	Year of the index date.	Remove
81	Demographics Post Observation Time	Number of continuous days of observation time following the index date.	Remove
87	Procedure Occurrence Any Time Prior	One covariate per procedure in the procedure_occurrence table any time prior to index.	Remove
103	Visit Count Long Term	The number of visits observed in the long term window.	Remove

Figure 11.7: □□□□□□ - □□□□

Subgroup analyses

New subgroup

Female

Calculate subgroup analyses only

Female

having **all** ▾ of the following criteria:

+ Add criteria to group... ▾

with the following event criteria:

+ Add attribute... ▾

X with a gender of: **FEMALE** **Add** **Import** **Delete Criteria**

Figure 11.8: □□□□□□□□ - □□□□□□□□□□□□



11.7.2

Design	Executions	Utilities
<h2>Executions</h2>		
SYNPUF 1K	Generate	View latest result
SYNPUF 5%	Generate	View latest result

Figure 11.9: □□□□□□□□□ - CDM□□□□□

All Executions View
Reports View latest result

11.7.3

CONDITION / Condition Occurrence Long Term / stratified by Female												
Export Export comparison			Show 10 ▾ entries					Search: <input type="text" value="cardi"/>				
Covariate	Explore	Concept ID	Patients initiating first-line therapy for hypertension with > 1 yr follow-up				Patients initiating first-line therapy for hypertension with > 3 yr follow-up				Std diff ▾	
			Count		Pct		Female		Count			
			Count		Pct		Count		Pct			
Tachycardia	Explore ▾	444070	17,322	1.04%	9,042	1.18%	6,547	0.78%	3,530	0.90%	-0.0193	
Cardiomegaly	Explore ▾	314658	20,958	1.26%	8,007	1.04%	9,016	1.08%	3,465	0.89%	-0.0121	
Cardiac arrhythmia	Explore ▾	44784217	30,474	1.83%	13,221	1.72%	14,540	1.74%	6,318	1.62%	-0.0052	

Figure 11.10: □□□□□□□ - □□1□□□□□□□

11.10 365

© Springer-Verlag Berlin Heidelberg 2009

Exploring condition_occurrence during day -365 through 0 days relative to index: Cardiac arrhythmia							X
Cohort: Patients initiating first-line therapy for hypertension with >1 yr follow-up							
Export	Show 10 ▾ entries					Search: <input type="text"/>	
Relationship type	Distance	Concept name		All stratas		Female	
		Count	Pct	Count	Pct	Count	Pct
Explore Ancestor	4	Disorder by body site		32	0.00%	17	0.00%
Explore Ancestor	4	Finding of trunk structure		991	0.06%	605	0.08%
Explore Ancestor	3	Disorder of trunk		23	0.00%	14	0.00%
Explore Ancestor	3	Disorder of thorax		241	0.01%	104	0.01%
Explore Ancestor	3	Disorder of body system		4,135	0.25%	1,992	0.26%
Explore Ancestor	2	Disorder of cardiovascular system		12,979	0.78%	6,073	0.79%
Explore Ancestor	2	Disorder of mediastinum		138	0.01%	62	0.01%
Explore Ancestor	2	Disorder of body cavity		24	0.00%	10	0.00%
Explore Ancestor	1	Heart disease		4,691	0.28%	1,869	0.24%
Explore Selected	0	Cardiac arrhythmia		30,474	1.83%	13,221	1.72%

Showing 1 to 10 of 62 entries

Previous 1 2 3 4 5 6 7 Next

Figure 11.11: □□□□□□□ - □□□□□□□□□

CONDITION / Condition Occurrence Long Term / stratified by Female

Export	Export comparison	Show 10 ▾ entries	Patients initiating first-line therapy for hypertension with >1 yr follow-up				Patients initiating first-line therapy for hypertension with >3 yr follow-up				Std diff ▾	
Covariate	Explore	Concept ID	Female		Female		Female		Female			
			Count	Pct	Count	Pct	Count	Pct	Count	Pct		
Edema	Explore ▾	433595	32,243	1.94%	20,200	2.63%	15,173	1.81%	9,684	2.48%	-0.0066	

Showing 1 to 1 of 1 entries (filtered from 206 total entries)

Previous 1 Next

Figure 11.12: □□□□□□□ - □□□□□□□□□

explore

Exploring condition_occurrence during day -365 through 0 days relative to index: Edema							
Cohort: Patients initiating first-line therapy for hypertension with >1 yr follow-up							
Export		Show 10 ▾ entries	Search: and				
Relationship type	Distance	Concept name	All stratas		Female		
			Count	Pct	Count	Pct	
Explore Descendant	-2	Angioedema	2,605	0.16%	1,506	0.20%	

Figure 11.13: □□□□□□ - □□□□□□□□□□

DEMOGRAPHICS / Demographics Age

Strata		Patients initiating first-line therapy for hypertension with >1 yr follow-up				Patients initiating first-line therapy for hypertension with >3 yr follow-up				Std diff ▼
		Count	Avg	Std Dev	Median	Count	Avg	Std Dev	Median	
Female	768,180	49.39	9.78	51.00	390,693	49.01	9.03	51.00	-0.0291	
All stratas	1,661,604	48.96	10.00	50.00	837,459	48.64	9.26	50.00	-0.0232	

Figure 11.14: □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□

11.7.4

ATLAS Analysis New Feature Analysis 

██████████ ACE ██████████

Analysis

118 R

 OHDSI

Design

Criteria Custom

Analysis type:

Prevalence

Add Criteria feature

Ace inhibitor exposure after index

having all of the following criteria:

+ Add criteria to group...

with at least 1 using all occurrences of:

+ Add attribute...

a drug era of ACE inhibitors

where event starts between 1 days After and All days After index start date add additional constraint

Delete Criteria

allow events from outside observation period

Figure 11.15: ATLAS□□□□□□□□□

DRUG / Ace inhibitor exposure after index / stratified by Female

			Patients initiating first-line therapy for hypertension with >1 yr follow-up				Patients initiating first-line therapy for hypertension with >3 yr follow-up				Std diff	
Covariate	Explore	Concept ID	Count	Pct	Female		Count	Pct	Female			
					Count	Pct			Count	Pct		
Ace inhibitor exposure after index	Explore	0	686,034	41.29%	289,215	17.41%	426,280	50.90%	182,219	21.76%	0.1001	

Showing 1 to 1 of 1 entries

Previous 1 Next

Figure 11.16: □□□□□□□□□□□□□

- Feature Extraction
- Cohort Definition
- Cohort Selection

Feature Extraction → 2

11.8.1

→ Chapter 10 B.6 → Appendix B → scratch.my_cohorts

11.8.2

→ R → Feature Extraction → Database Connector

```
library(FeatureExtraction)
connDetails <- createConnectionDetails(dbms = "postgresql",
                                         server = "localhost/ohdsi",
                                         user = "joe",
                                         password = "supersecret")

cdmDbSchema <- "my_cdm_data"
cohortsDbSchema <- "scratch"
cohortsDbTable <- "my_cohorts"
cdmVersion <- "5"
```

→ 4 → cdmDbSchema → cohortsDbSchema → cohortsDbTable → CDM
 SQL Server → cohortsDbSchema → cohortsDbTable → cohortsDbTable →
 ← "my_cdm_data.dbo"

11.8.3

createCovariateSettings →

```
settings <- createCovariateSettings(
  useDemographicsGender = TRUE,
  useDemographicsAgeGroup = TRUE,
  useConditionOccurrenceAnyTimePrior = TRUE)
```

→ 5 →

- 365
- 180
- 30

```
settings <- createCovariateSettings(useConditionEraLongTerm = TRUE,  
                                     useConditionEraShortTerm = TRUE,  
                                     useDrugEraLongTerm = TRUE,  
                                     useDrugEraShortTerm = TRUE,  
                                     longTermStartDays = -180,  
                                     shortTermStartDays = -14,  
                                     endDays = -1)
```

180 14
ID

```
settings <- createCovariateSettings(useConditionEraLongTerm = TRUE,  
                                     useConditionEraShortTerm = TRUE,  
                                     useDrugEraLongTerm = TRUE,  
                                     useDrugEraShortTerm = TRUE,  
                                     longTermStartDays = -180,  
                                     shortTermStartDays = -14,  
                                     endDays = -1,  
                                     excludedCovariateConceptIds = 1124300,  
                                     addDescendantsToExclude = TRUE,  
                                     aggregated = TRUE)
```



aggregated = TRUE Feature Extraction

11.8.4

```
covariateSettings <- createDefaultCovariateSettings()  
  
covariateData2 <-getDbCovariateData(  
  connectionDetails = connectionDetails,  
  cdmDatabaseSchema = cdmDatabaseSchema,  
  cohortDatabaseSchema = resultsDatabaseSchema,  
  cohortTable = "cohorts_of_interest",  
  cohortId = 1,  
  covariateSettings = covariateSettings,  
  aggregated = TRUE)  
  
summary(covariateData2)
```

```
## CovariateData Object Summary  
##  
## Number of Covariates: 41330  
## Number of Non-Zero Covariate Values: 41330
```

11.8.5

```
covariateData2$covariates  
covariateData2$covariatesContinuous
```

11.8.6

Feature Extraction // ohdsi.github.io/FeatureExtraction/

11.9 ATLAS

Hripcsak7329 ATLAS Cohort
Pathways



Figure 11.17: □□□□□□□□□□□□□□□□□□□□□□□□

ATLAS Cohort Pathways

11.9.1

Cohort Pathway is defined as the process of generating an aggregated sequence of transitions between the Event Cohorts among those people in the Target Cohorts.

Target Cohorts

Each of the Target Cohorts will be analyzed for the pathways through the event cohorts.

Import			
Show 10 ▾ entries		Search: <input type="text"/>	
ID	Name	Actions	Actions
10447	Patients initiating first-line therapy for hypertension with >1 yr follow-up	Edit cohort	Remove
10448	Patients initiating first-line therapy for hypertension with >3 yr follow-up	Edit cohort	Remove

Showing 1 to 2 of 2 entries

Previous 1 Next

Figure 11.18: □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□

ACE

B.8-B.16 Import

- 2: 2
- 3: 3
- 4: 4

11.9.2

11.9.3

11.10 ATLAS

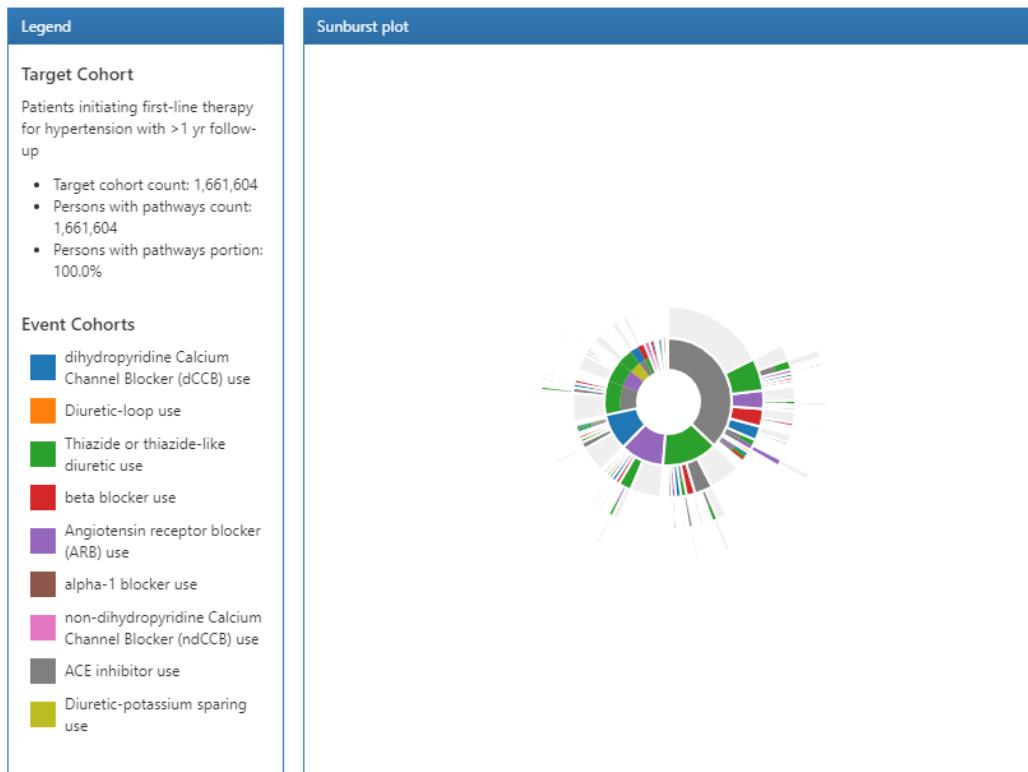
(ACEi) (THZ) (An-gioedema) (AMI) (TAR) (ACEi THZ)

Event Cohorts

Each Event Cohort defines the step in a pathway that may occur for a person in the Target Cohort.

Import

ID	Name		
9174	ACE inhibitor use	Edit cohort	Remove
9175	Angiotensin receptor blocker (ARB) use	Edit cohort	Remove
9176	Thiazide or thiazide-like diuretic use	Edit cohort	Remove
9177	dihydropyridine Calcium Channel Blocker (dCCB) use	Edit cohort	Remove
9178	non-dihydropyridine Calcium Channel Blocker (ndCCB) use	Edit cohort	Remove
9179	beta blocker use	Edit cohort	Remove
9180	Diuretic-loop use	Edit cohort	Remove
9181	Diuretic-potassium sparing use	Edit cohort	Remove
9182	alpha-1 blocker use	Edit cohort	Remove



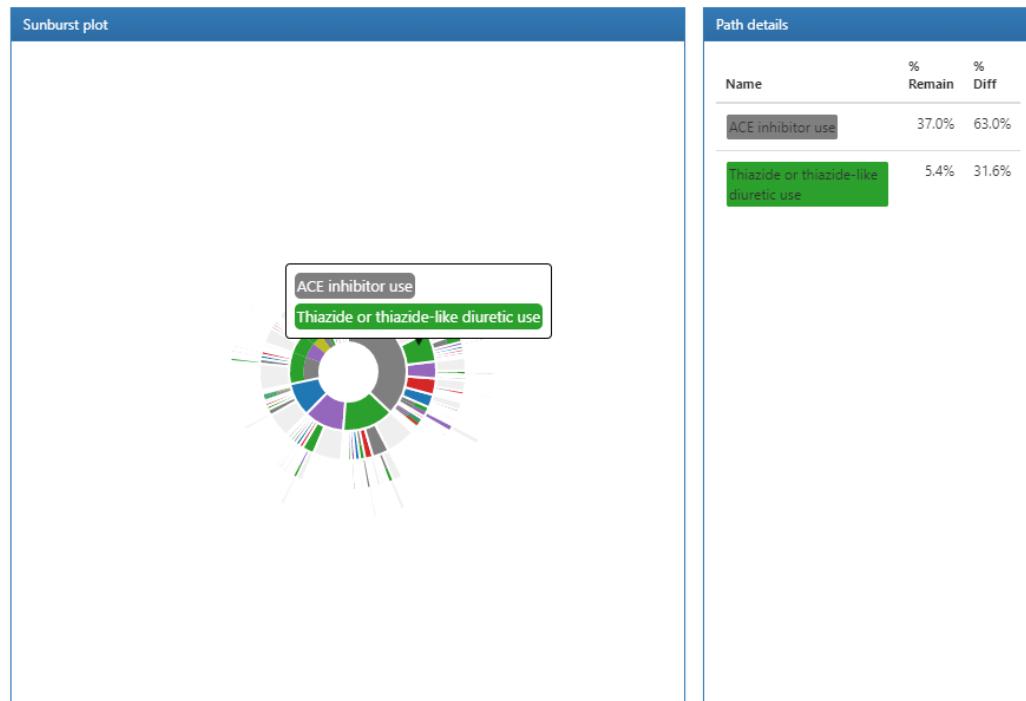


Figure 11.21: □□□□□□□□□□□□□□□□□□□□□□□□

ATLAS Incidence Rates

11.10.1

□□□□□□□□□□□□□□□□□ ATLAS □□□□□□□□□□□□□□□ (Chapter 10
□□□)□□□□□□□□□□□□□□□ (Appendix B.2, B.5) □□□□□□□□□□□□□
(Appendix B.4, B.3, B.9) □□□□□□□□□□□□□□□

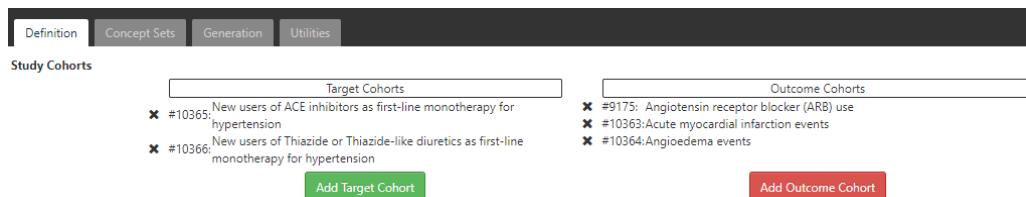


Figure 11.22: □□□□□□□□□□□□□□□□□□□□□□□□

Time At Risk

Time at risk defines the time window relative to the cohort start or end date with an offset to consider the person 'at risk' of the outcome.

- Time at risk starts with **start date** ▾ plus **1** ▾ days.
 - Time at risk ends with **end date** ▾ plus **0** ▾ days.

No study window defined.

Add Study Window

Figure 11.23: □□□□□□□□□□□□□□□□□□□□□□□□□□□□□

THz

ATLAS :

Stratify Criteria: You can provide optional stratification criteria to the analysis that will divide the population into unique groups based on their satisfied criteria.

New stratify criteria	Gender = Female	Copy	Delete
<p>1. Gender = Female</p> <p>enter an inclusion rule description</p> <p>having <input style="border: 1px solid #ccc; padding: 2px 10px;" type="button" value="all"/> of the following criteria:</p> <p>+ Add criteria to group...</p> <p>with the following event criteria:</p> <p>+ Add attribute...</p> <p>✖ with a gender of: <input style="border: 1px solid #ccc; padding: 2px 10px;" type="button" value="FEMALE"/> Add Import</p> <p>Delete Criteria</p>			

Figure 11.24: □□□□□□□□□□□□□□

Chapter 11

11.10.2

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[]

11.10.3

• Reports

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1000

1000 □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □

Select sources X

Filter:

Name	▲
<input type="checkbox"/> SYNPUF 1K	
<input checked="" type="checkbox"/> SYNPUF 5%	

Previous 1 Next

Figure 11.25: □ □ □ □ □ □ □ □

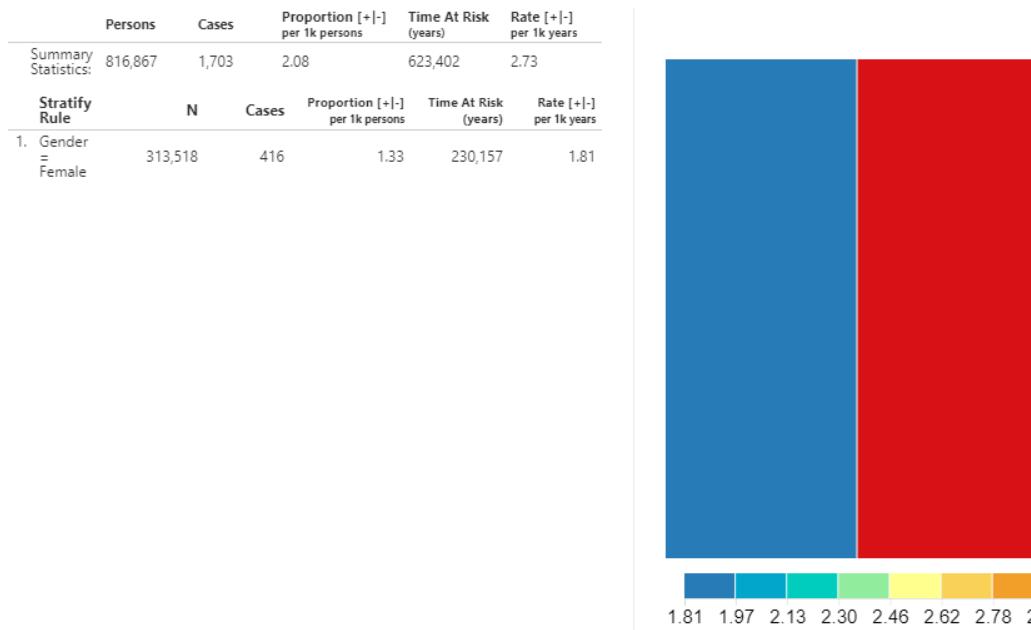


Figure 11.26: □ □ □ □ □ □ □ □ - AMI □ □ □ □ □ □ □ □ □ □ □ ACEi □ □ □ □

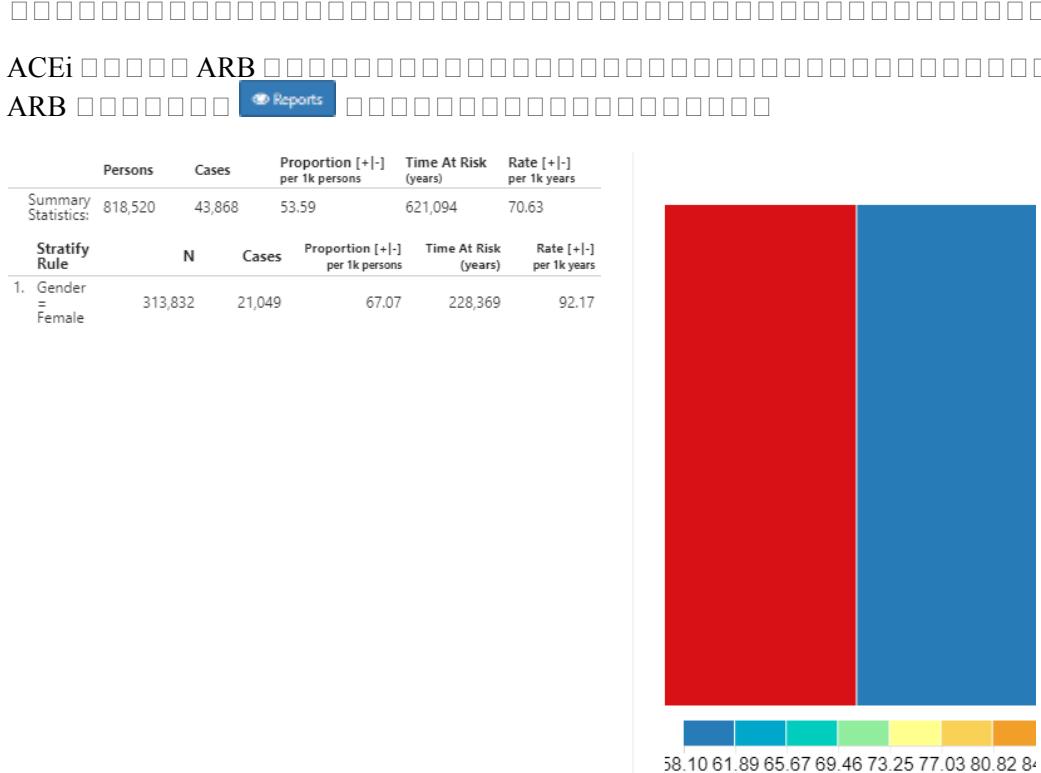


Figure 11.27: □□□ - ACEi □□□□ ARB □□□□□□□□□□ ACEi □□□□

11.11



- OHDSI
 -
 - ATLAS OHDSI
 - ATLAS

11.12



ATLAS <http://atlas-demo.ohdsi.org>

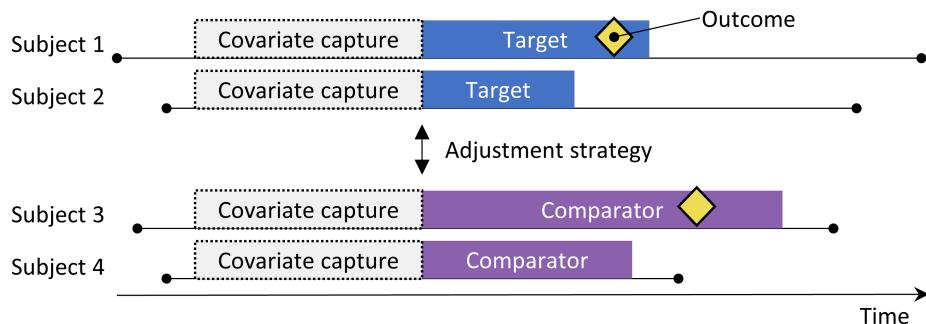
□□□□□□□□□ Appendix E.7 □□□□□□□□□

Chapter 12

12 — —

 <img alt="PubMed icon" data-bbox="6230 881 6

12.1



□: □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □



12.1.1

PS Jan 0.4
Jun

12.1.2

PS OHDSI et al., 2018,) et al., 2013,) Cyclops



12.1.3

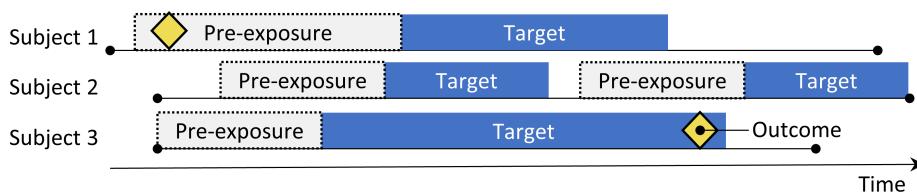
12.1.4

$$\ln\left(\frac{F}{1-F}\right) = \ln\left(\frac{S}{1-S}\right) - \ln\left(\frac{P}{1-P}\right)$$

12.1.5

PS 12.11 2001,)

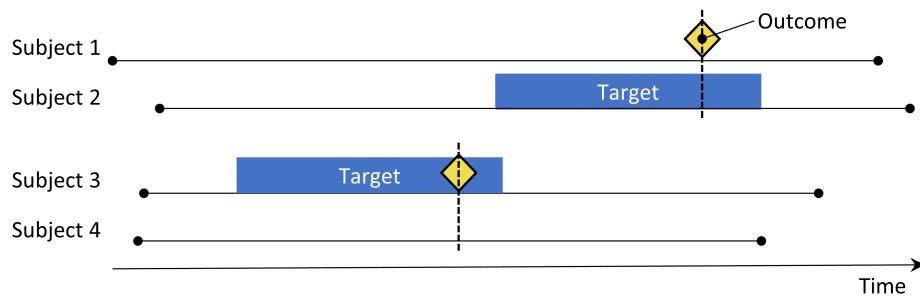
12.2



□□□□□□□□□SCC□□□□□(Ryan et al., 2013a,) □□□□
12.2 □□□4□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□

12.3

Table 12.2: □□□□□□□□□□□□□□□□□□□□□□□□□□□□□



12.4

12.5

SCCS (Farrington, 1995; Whitaker et al., 2006,
)
SCCS

Table 12.3: □□□□□□□□□□□□□□□□□□

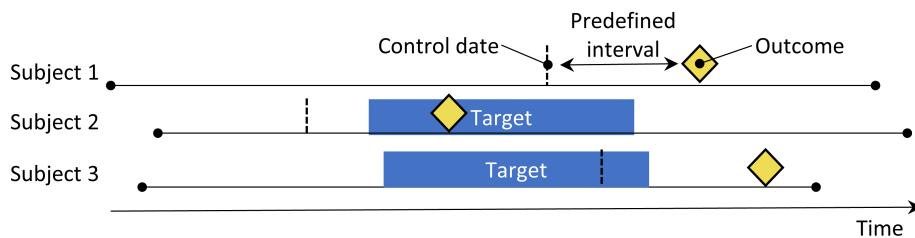


Table 12.4: □□□□□□□□□□□□□□□□□□□□□□

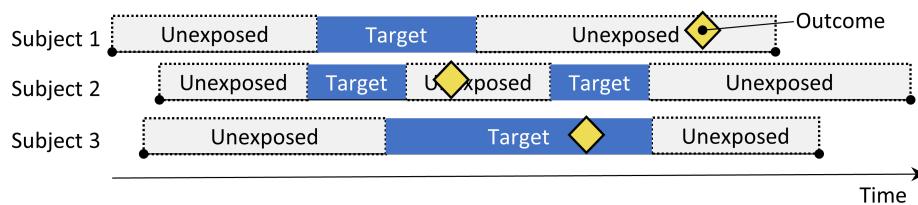
SCCS et al., 2013,)

12.6

12.6.1

ACE (Wang et al., 2002,) and Black, 1997) (Wang et al., 2012,) ACEi β (Wang et al., 2010; Toh et al., 2012,) β (Whelton et al., 2018,) THZ

Table 12.5: □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□



ACE

12.6.2

ACEi THZ

12.6.3

ER 7

12.6.4

12.6.5

PS

12.6.6

□□□ □

□□□□□□□□ □□□□□□□□□□□□ ACE □□□□□□□□
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12.6.7

MethodValidity
@ref

12.7 ATLAS

ATLAS
Estimation
G

3

12.7.1

1 Appendix B.2 Appendix @ref(ThiazidesMono
@ref(Angioedema
Appendix @ref(Ami
12.6

 Comparison
Add or update the target, comparator, outcome(s) cohorts and negative control outcomes

Choose your target cohort:

New users of ACE inhibitors as first-line monotherapy for hypertension
 

Choose your comparator cohort:

New users of Thiazide-like diuretics as first-line monotherapy for hypertension
 

Choose your outcome cohorts:

Add Outcome

Show 10 entries		Search: <input type="text"/>	
ID	Name	Edit cohort	Remove
1770712	Angioedema outcome	Edit cohort	Remove
1770713	Acute myocardial infarction outcome	Edit cohort	Remove

Showing 1 to 2 of 2 entries

Previous 1 Next

Figure 12.6: ATLAS

MethodValidity
@ref

18

12.7

Negative controls for ACEi and THZ									Optimize							
Concept Set Expression		Included Concepts 75		Included Source Codes		Explore Evidence		Export		Compare						
Show 25 ▾ entries						Search: <input type="text"/>										
Showing 1 to 25 of 75 entries																
	Concept Id	Concept Code	Concept Name	Domain	Standard Concept Caption	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	Previous 1	2	3	Next				
	72748	74779009	Strain of rotator cuff capsule	Condition	Standard	<input checked="" type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>								
	73241	197210001	Anal and rectal polyp	Condition	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>								
	73560	55260003	Calcaneal spur	Condition	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>								
	75911	65358001	Acquired hallux valgus	Condition	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>								
	76786	63643000	Derangement of knee	Condition	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>								

Figure 12.7: □□□□□□□□□□□□

Concept Set #1798551															
Concepts to exclude for ACEi and THZ															
Concept Set Expression		Included Concepts		Included Source Codes		Explore Evidence		Export	Compare						
		38225													
Show 25 ▾ entries									Search: <input type="text"/>						
Showing 1 to 14 of 14 entries									Previous 1 Next						
Concept Id	Concept Code	Concept Name	Domain	Standard Concept Caption	Exclude	Descendants	Mapped								
1342439	38454	trandolapril	Drug	Standard	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>								
1334456	35296	Ramipril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>								
1331235	35208	quinapril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>								
1373225	54552	Perindopril	Drug	Standard	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>								
1310756	30131	moexipril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>								

Figure 12.8: □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□

Figure 12.9

Negative control concept set:

Negative controls for ACEi and THZ



Covariate selection

Concepts to **include** when constructing the covariates to be used in this study. (Leave blank if you want to include every concept).*



* Concepts defined here are combined with those defined in the Analysis settings section.

Concepts to **exclude** when constructing the covariates to be used in this study.*

Concepts to exclude for ACEi and THZ



* Concepts defined here are combined with those defined in the Analysis settings section.

Figure 12.9: A screenshot of a software interface showing a negative control concept set for 'ACEi and THZ'. It includes sections for 'Covariate selection' (with an 'include' field) and 'Exclude' (with an 'ACEi and THZ' field). Both sections have blue and red file operation buttons at the top right.

12.7.2

R

R

R
R
R
R

12.8 R

R R ATLAS R R
R

CohortMethod CohortMethod CDM
12.8.6 AMI

12.8.1

(10) (B.2) (B.5)
(B.4) ACEi THZ
1 2 3 scratch.my_cohorts

12.8.2

R CohortMethod DatabaseConnector

```
library(CohortMethod)
connDetails <- createConnectionDetails(dbms = "postgresql",
                                         server = "localhost/ohdsi",
                                         user = "joe",
                                         password = "supersecret")

cdmDbSchema <- "my_cdm_data"
cohortDbSchema <- "scratch"
cohortTable <- "my_cohorts"
cdmVersion <- "5"
```

```
cdmDbSchema cohortDbSchema cohortTable CDM  
SQL Server cdm  
<- "my_cdm_data.dbo"
```

CohortMethod

```
## CohortMethodData  
##  
##      ID 1  
##      ID 2  
##      ID(s) 3
```

CohortMethod createDefault
(12.1) 2

cohortsMethodData
(8.4.2)□□□□□□□

summary()

```
summary(cmData)
```

```
## CohortMethodData
##
##      ID 1
##      ID 2
##      ID(s) 3
##
##      67166
##      35333
##
##      980          891
##  3
##
##      58349
##      24484665
```

cohortMethodData ━━━━━━

```
loadCohortMethodData()
```

1 CohortMethod

12.8.3

```
getAttritionTable(studyPop)
```

```
##                                     ...
## 1                               67212      35379 ...
## 2           67166      35333 ...
## 3                               67061      35238 ...
## 4     1       66780      35086 ...
```

12.8.4

```
getDbcohortMethodData()
```

```
ps <- createPs(cohortMethodData = cmData, population = studyPop)
```

createPs Cyclops

```
matchedPop <- matchOnPs(population = ps, caliper = 0.2,  
                           caliperScale = "standardized logit", maxRatio = 100)
```

□□□□□PS□trimByPs□trimByPsToEquipoise□□□□stratifyByPs□□□□□□□□□□□□□□

12.8.5

```
outcomeModel <- fitOutcomeModel(population = matchedPop,
                                   modelType = "cox",
                                   stratified = TRUE)

outcomeModel

##      cox
##    TRUE
##   FALSE
##   FALSE
##     OK
##
##                95%  95% logRr seLogRr
## 4.3203    2.4531   8.0771 1.4633  0.304
```

12.8.6

```
#  
ois <- c(3, 4) #      AMI  
  
#  
ncs <- c(434165, 436409, 199192, 4088290, 4092879, 44783954, 75911, 137951, 77965,  
       376707, 4103640, 73241, 133655, 73560, 434327, 4213540, 140842, 81378,  
       432303, 4201390, 46269889, 134438, 78619, 201606, 76786, 4115402,  
       45757370, 433111, 433527, 4170770, 4092896, 259995, 40481632, 4166231,  
       433577, 4231770, 440329, 4012570, 4012934, 441788, 4201717, 374375,  
       4344500, 139099, 444132, 196168, 432593, 434203, 438329, 195873, 4083487,  
       4103703, 4209423, 377572, 40480893, 136368, 140648, 438130, 4091513,  
       4202045, 373478, 46286594, 439790, 81634, 380706, 141932, 36713918,  
       443172, 81151, 72748, 378427, 437264, 194083, 140641, 440193, 4115367)  
  
tcos <- create  
  
##      {#studyOutputs}
```

ATLAS R

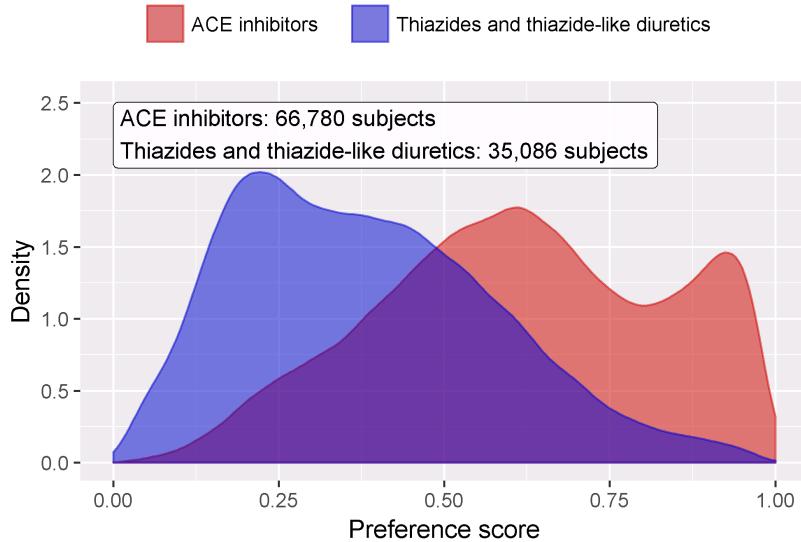


Figure 12.10: □□□□□□□□

-1.42 -30 0 :

-1.11 0 0 :

0.68 : 05-09

0.64 □□□□□-365□□□0□□□□□□□□□: □□

0.63 □□□□□-30□□□0□□□□□□□□□□: □□

0.55 -365 0 :

0.52 (□□)

0.50 □ □ :

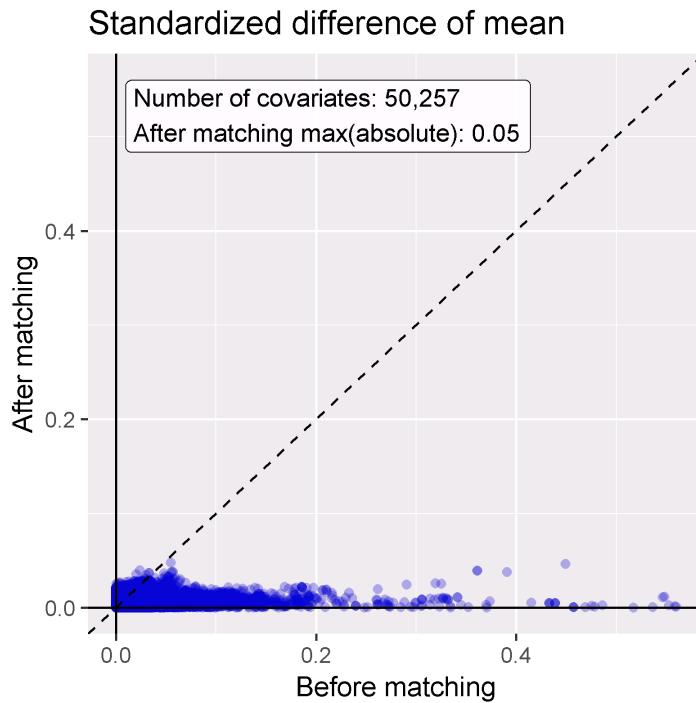
!

12.8.7

PS 2 12.11 0.1

12.8.8

12.12 drawAttritionDiagram



14

12.9

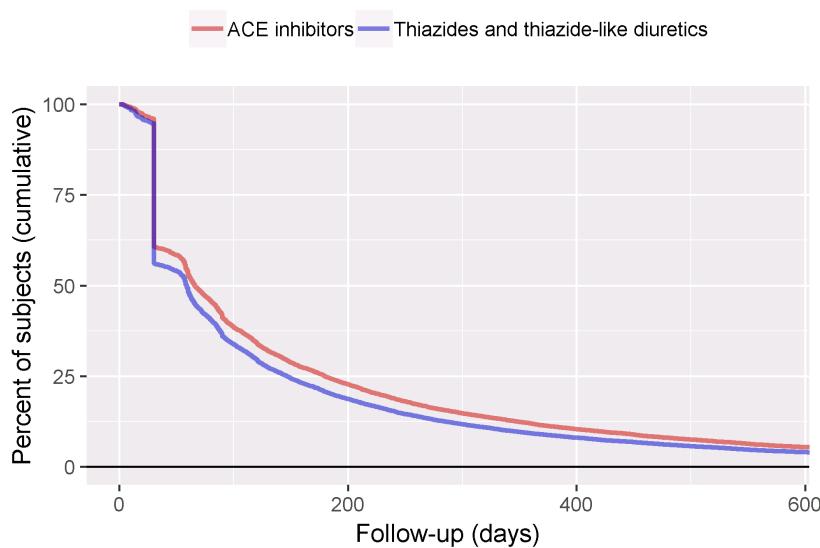
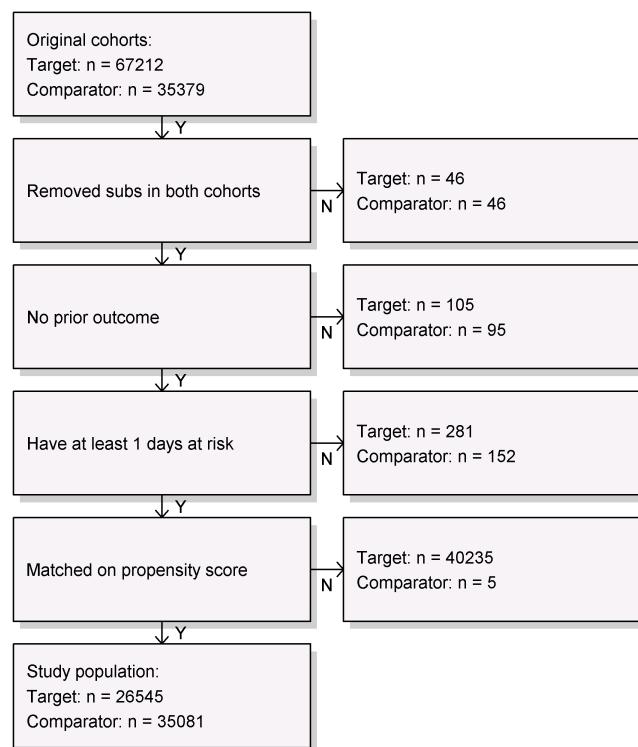


12.10

100

R R-Studio Java 8.4.5

```
install.packages(c("SqlRender", "DatabaseConnector", "remotes"))
remotes::install_github("ohdsi/Eunomia", ref = "v1.0.0")
```



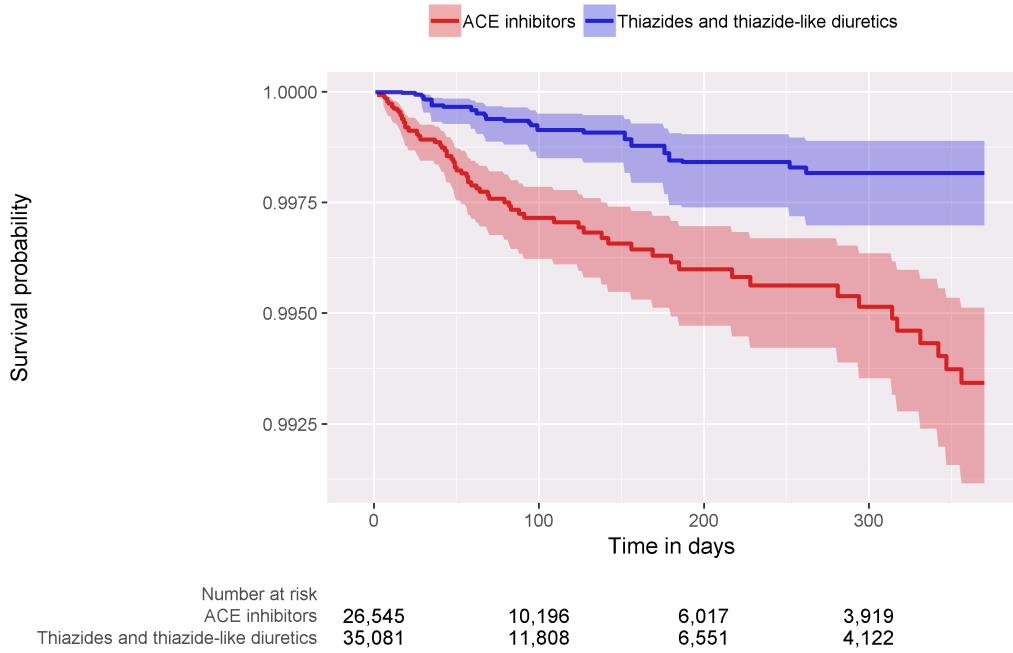


Figure 12.14: □□□□-□□□□□□□□□□

```
remotes::install_github("ohdsi/CohortMethod")
```

Eunomia R CDM

```
connectionDetails <- Eunomia::getEunomiaConnectionDetails()
```

CDM main Eu

```
Eunomia:::createCohorts(connectionDetails)
```

1

□□□□□□□□□□□□□□□□□ COHORT DEFINITION ID □1□□□□□□□□□□□□□□□□□

□ □ 12.5. 5 □ □ □ □ □ □ PS □

E.8

Chapter 13

13 — —

Lead: Peter Rijnbeek & Jenna Reps

OHDSI CDM

QUDSI et al. / J Child Psychol Psychiatr 2018; 59(10): 1039–1047

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[□□□□□□□□□□□□] □□□□□□□□□□□□ [□□□□□□□□□□□□] □□□□□□

□: □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □

¹<https://www.equator-network.org/reporting-guidelines/tripod-statement/>

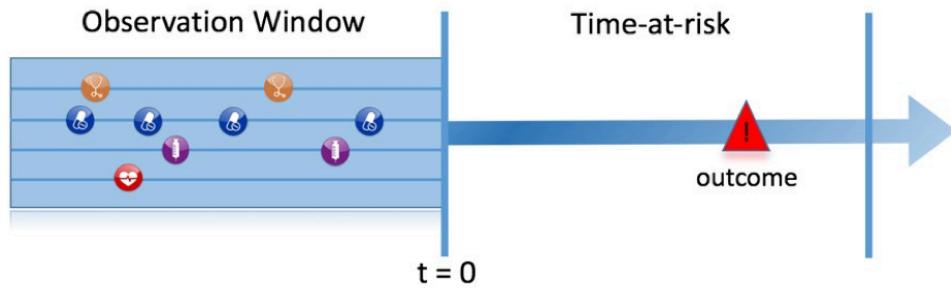


Figure 13.1: □□□□□

A horizontal row of 30 empty square boxes, each with a thin black border, intended for individual names or responses.

13.2

13.2.1

13.2.1 2 COHORT ID 1

□□□□□COHORT□□□□□□□□□□□COHORT_END_DATE□□□□□□□□□

COHORT_DEFINITION_ID	SUBJECT_ID	COHORT_START_DATE
1	1	2000-06-01
1	2	2001-06-01
2	2	2001-07-01

□□□□□ CONDITION OCCURRENCE □□□□□□□□□□□ 3 □□□□□□□□□□

PERSON_ID	CONDITION_CONCEPT_ID	CONDITION_START_DATE
1	320128	2000-10-01
2	320128	2001-05-01

13.2.2 vs

13.3

A horizontal row of 20 empty square boxes, intended for students to write their answers in a grid format.

PatientLevelPrediction

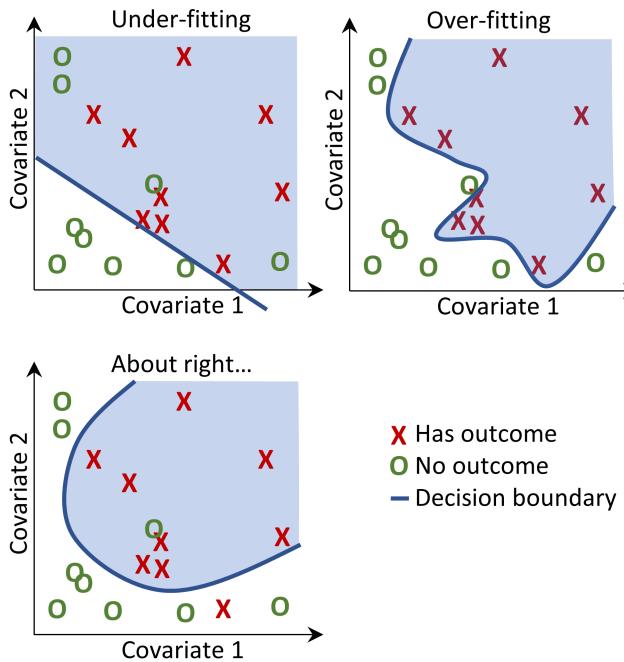


Figure 13.2: Decision boundary.

13.3.1

The diagram consists of two horizontal rows of rectangular boxes. The top row contains 11 boxes: four filled boxes on the left, two empty boxes in the center, and four filled boxes on the right. The bottom row contains 10 boxes: three filled boxes on the left, five empty boxes in the center, and one large box on the right labeled '0.1'.

13.3.2

Table 13.5: □□□

earlyStopRound 25

13.3.3

Table 13.6: □□□□□□□□□□□□□□□□□□□□□□□□

<input type="checkbox"/>					
maxDepth	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	4,10,17
mtries	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	-1 = <input type="checkbox"/> <input type="checkbox"/> <input type="checkbox"/> <input type="checkbox"/> <input type="checkbox"/> <input type="checkbox"/> ,5,20
ntrees	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	500	

13.3.4 K-

K-nearest neighbors (KNN)

Table 13.7: K-□□□□□□□□□□□□□□□□

<input type="checkbox"/>						
k		<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	1000	

13.3.5

13.3.6 AdaBoost

AdaBoost
AdaboostClassifier

Table 13.8: AdaBoost

13.3.7

DecisionTreeClassifier

Table 13.9: DecisionTreeClassifier

classWeight	“Balance” or “None”	None
maxDepth	10	
minImpuritySplit	10^-7	
minSamplesLeaf	10	
minSamplesSplit	2	

13.3.8

Table 13.10: RidgeClassifier

alpha	1.2	0.00001
size	4	

13.3.9

RidgeClassifier

13.3.10

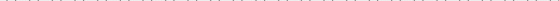
Patient-level prediction

13.4

13.4.1

Patient-level prediction



13.3 

A horizontal row of 20 empty square boxes, each with a thin black border, intended for children to practice writing their names.

□□□□□□□□□□□□□□□□**2**□□□□□□□□

13.4.2

10 of 10

13.11 0.5 13.7 10 0.5

Table 13.11: □□□□□□□□□□□□□□

ID	0.5	0.5	0.5	0.5
1	0.8	1	1	TP
2	0.1	0	0	TN
3	0.7	1	0	FP
4	0	0	0	TN
5	0.05	0	0	TN
6	0.1	0	0	TN
7	0.9	1	1	TP
8	0.2	0	1	FN
9	0.3	0	0	TN
10	0.5	1	0	FP

TP

A horizontal row of 20 empty rectangular boxes, intended for handwritten responses or drawings.

- $\square\square$: $(TP + TN) / (TP + TN + FP + FN)$

- $\text{Sensitivity} = TP / (TP + FN)$
- $\text{Specificity} = TN / (TN + FP)$
- $\text{Precision} = TP / (TP + FP)$

$FN = \text{False Negatives}$

$\text{AUC} = \text{Area Under the Curve}$

- $\text{AUC} = \text{ROC Area} = \text{Probability of a randomly selected case having a higher predicted risk than a randomly selected control}$

13.11 $\text{AUC} = 0.5 \rightarrow \text{No discrimination}$

$\text{AUC} = 0.5 \rightarrow \text{Poor discrimination}$

13.3 $\text{AUC} = 1.0 \rightarrow \text{Perfect discrimination}$

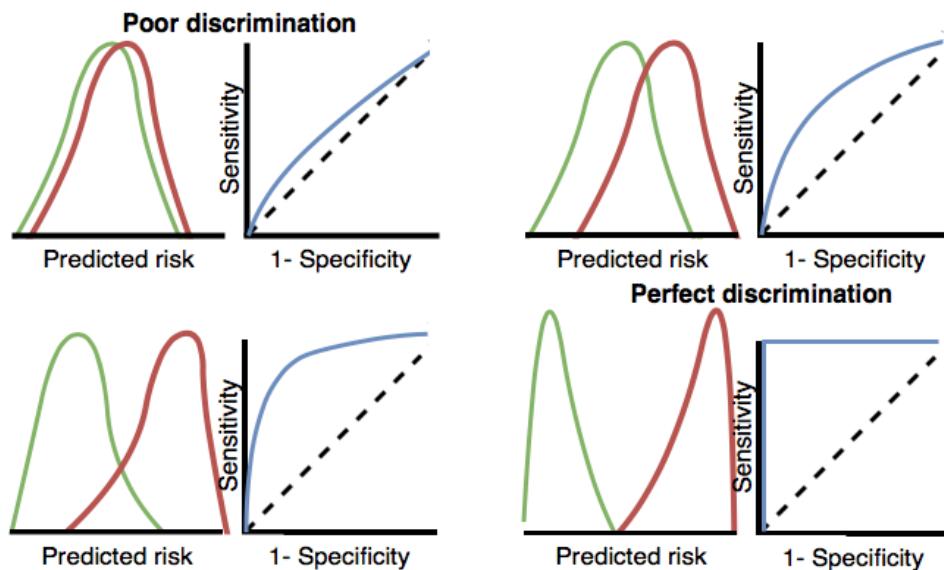


Figure 13.3: $\text{AUC} = \text{ROC Area}$

$\text{AUC} = \text{Area Under the ROC Curve}$

$\text{AUPRC} = \text{Area Under the Precision-Recall Curve}$

$\text{Calibration Curves}$

$\text{y} = \text{True Positive Rate}$

13.12 $\text{Calibration Curves}$

$\text{Calibration Curves}$

13.5

A horizontal row of twelve empty square boxes, intended for children to write their names in, likely as part of a classroom activity.

13.1

13.5.1

□ □ □ □ □ □ □ ACE □

(Byrd et al., 2006,)□ □

et al., 2013,)□ □ □ □ □

et al., 2013; Thompson

et al., 2006,)□□□□□

et al., 2004.)

and O'Mara 1996) □

A horizontal row of twelve empty square boxes, intended for children to write their names in, likely as part of a classroom activity.

ACE

13.5.2

- □□□□□□□□□□
 - □□□□□□□□□□
 - □□□□□□□□□□
 - □□□□□□□□□□
□□□□□□□□□□
 - □□□□□□□□□□

13.5.3

13.3

Boosting Machines □ G

□ □ □ □ □ □ □ □ □ □ □

13.5.4

- 25 -

13.5.5

□□□□□ 13.5.5□□

□ : □ □ □ □ □ □ □ □ □ □ □ □ □

13.6 ATLAS

ATLAS Prediction 4

13.6.1

13.6.2

A horizontal row of 24 small, identical rectangular boxes arranged in a single line.

10

Model Settings ATLAS

 Prediction Problem Settings

Target Cohorts

Show entries Filter:

Remove **Name**

 New users of ACE inhibitors as first-line monotherapy for hypertension
--

Showing 1 to 1 of 1 entries Previous Next

Outcome Cohorts

Show entries Filter:

Remove **Name**

 Angioedema outcome
--

Showing 1 to 1 of 1 entries Previous Next

Figure 13.4: □□□□□□.

5 / 5

Add Covariate Settings

concepts do you want to include in baseline covariates in the patient-level prediction model? (Leave blank if you want to include everything)

descendant concepts be added to the list of included concepts? yes
concepts do you want to exclude in baseline covariates in the patient-level prediction

13.6□□□□□□□□

A horizontal row of 20 empty white rectangular boxes, likely used for input fields or placeholder text in a form.

- □□: □□□□□□□□□□□□□□□□□□
 - □□: □□□□□□□□□
 - □□□□□□: 5□□□□□□□□□□□□□0-4□5-9□...□95+□
 - □□: □□□□□□□□□□□
 - □□: □□□□□□□□□□□
 - □□□□□□□□: □□□□□□□□□□□□□□□□□□
 - □□□□□□□□: □□□□□□□□□□□□□□□□□□
 - □□□□□□□□: [□□□□□□□□□□□□] □□□□□□□□□□□
 - □□□□□□□□: [□□□□□□□□□□□□] □□□□□□□□□□□
 - □□□□□□□□: □□□□□□□□□□□□
 - □□□□□□□□□□: [□□□□□□□□□□□□] □□□□□□□□□□□□□□□□□□□□

Gradient Boosting Machine Model Settings
Use the options below to edit the model settings

The boosting learn rate (default = 0.01,0.1):

Boosting learn rate	Action
0.001	Remove
0.01	Remove
0.1	Remove
0.9	Remove

[Add](#) [Reset to default](#)

Maximum number of interactions - a large value will lead to slow model training (default = 4,6,17):

Maximum number of interactions	Action
4	Remove
7	Remove
10	Remove

[Add](#) [Reset to default](#)

The minimum number of rows required at each end node of the tree (default = 20):

Minimum number of rows	Action
20	Remove

[Add](#) Using default

The number of trees to build (default = 10,100):

Trees to build	Action
5000	Remove

[Add](#) [Reset to default](#)

The number of computer threads to use (how many cores do you have?) (default = 20):

20	Using default
----	---------------

Figure 13.5: □ □ □ □ □ □ □ □ □ □ □ □

What concepts do you want to include in baseline covariates in the propensity score model? (Leave blank if you want to include everything)

Save
Cancel

Should descendant concepts be added to the list of included concepts?

 No ▾

What concepts do you want to exclude in baseline covariates in the propensity score model? (Leave blank if you want to include everything)

Save
Cancel

Should descendant concepts be added to the list of included concepts?

 No ▾

A comma delimited list of covariate IDs that should be restricted to:

Figure 13.6: Configuration of baseline covariates.

13.7 Configuration of baseline covariates.

Select Covariates

	Gender	Age	Age Groups	Race	Ethnicity	Index Year	Index Month	Prior Observation Time	Post Observation Time	Time In Cohort	Index Year & Month
Demographics	<input checked="" type="checkbox"/>	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>

Figure 13.7: Configuration of select covariates.

13.8 Configuration of select covariates.

- Age: [-365, 0]
- [-365, -180]
- [-365, -30]

13.8 Configuration of time bound covariates.

Time bound covariates

Set the time windows for the time bound covariates in days relative to the cohort index

	Any Time Prior	Long Term	Medium Term	Short Term	End Days
Time Windows	All Time	-365	-180	-30	0

Figure 13.8: Configuration of time bound covariates.

13.9 Configuration of time bound covariates.

- Condition: ID
- Condition group: ID
- Drug: ID
- Drug group: ID

Set the time bound era covariates

Domain	Any Time Prior	Long Term (-365 days)	Medium Term (-180 days)	Short Term (-30 days)	Overlapping	Era Start		
						Long Term (-365 days)	Medium Term (-180 days)	Short Term (-30 days)
Condition	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Condition Group	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Drug	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Drug Group	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>

Figure 13.9: Set the time bound era covariates.

Set the time bound era covariates

- Condition: ID
- Condition Primary Inpatient: ID
- Drug: ID
- Procedure: ID
- Measurement: ID

13.7 R

ATLAS R PatientLevelPrediction CDM

13.7.1

10 Appendix (Appendix B.1) (Appendix B.4) ACE ID 1 ID 2

13.7.2

R PatientLevelPrediction DatabaseConnector

```
library(PatientLevelPrediction)
connDetails <- createConnectionDetails(dbms = "postgresql",
                                         server = "localhost/ohdsi",
                                         user = "joe",
                                         password = "supersecret")

cdmDbSchema <- "my_cdm_data"
cohortsDbSchema <- "scratch"
```

```
cohortsDbTable <- "my_cohorts"  
cdmVersion <- "5"
```

```
4 cdmDbSchema cohortsDbSchema cohortsDbTable CDM  
SQL Server  
<- "my_cdm_data.dbo"
```

```
sql <- paste("SELECT cohort_definition_id, COUNT(*) AS count",  
"FROM @cohortsDbSchema.cohortsDbTable",  
"GROUP BY cohort_definition_id")  
conn <- connect(connDetails)  
renderTranslateQuerySql(connection = conn,  
                         sql = sql,  
                         cohortsDbSchema = cohortsDbSchema,  
                         cohortsDbTable = cohortsDbTable)
```

```
##   cohort_definition_id  count  
## 1                      1 527616  
## 2                      2  3201
```

PatientLevelPrediction Features

```
covariateSettings <- createCovariateSettings(  
  useDemographicsGender = TRUE,  
  useDemographicsAge = TRUE,  
  useConditionGroupEraLongTerm = TRUE,  
  useConditionGroupEraAnyTimePrior = TRUE,  
  useDrugGroupEraLongTerm = TRUE,  
  useDrugGroupEraAnyTimePrior = TRUE,  
  useVisitConceptCountLongTerm = TRUE,  
  longTermStartDays = -365,  
  endDays = -1)
```

```
getPlpData
```

```
plpData <- getPlpData(connectionDetails = connDetails,  
                        cdmDatabaseSchema = cdmDbSchema,  
                        cohortDatabaseSchema = cohortsDbSchema,  
                        cohortTable = cohortsDbSchema,  
                        cohortId = 1,  
                        covariateSettings = covariateSettings,  
                        outcomeDatabaseSchema = cohortsDbSchema,  
                        outcomeTable = cohortsDbSchema,  
                        outcomeIds = 2,
```

```
    sampleSize = 10000  
)
```

getPlpData PatientLevelPrediction
plpData

```
savePlpData(plpData, "angio_in_ace_data")
```

loadPlpData()

13.7.3

```
2  
= 30  
riskWindowEnd = 365  
= TRUE
```

13.7.4

Gradient Boosting Machine ntr

```
= c(100,200), maxDepth = 4, nntrees = 100, maxDepth = 4, nntrees = 200, maxDepth = 4
```

runPlp(plpData, testSplit = 0.25, testFraction = 75%)

```
gbmResults <- runPlp(population = population,
                      plpData = plpData,
                      modelSettings = gbmModel,
                      testSplit = 'person',
                      testFraction = 0.25,
                      nfold = 2,
                      splitSeed = 1234)
```

R xgboost 75%

runPlp(plpData, plpResults, plpPlots, evaluation)

□□□□□□□□□□□

```
savePlpModel(gbmResults$model, dirPath = "model")
```

□□□□□□□□□□□

```
plpModel <- loadPlpModel("model")
```

□□□□□□□□□□□□□□□

```
savePlpResult(gbmResults, location = "gbmResults")
```

□□□□□□□□□□□□□□□

```
gbmResults <- loadPlpResult("gbmResults")
```

13.7.5

runPlp(gbmResults) Shiny
13.10

□□□□□□□□□□□□□□□

```
plotPlp(gbmResults, "plots")
```

□□□□□□□□□□□□□□□ Section 13.4.2

13.7.6

PLP Model

```
#  
plpModel <- loadPlpModel("model")  
  
# plpData  
plpData <- loadPlpData("newData")  
  
population <- createStudyPopulation(plpData = plpData,  
                                      outcomeId = 2,  
                                      washoutPeriod = 364,  
                                      firstExposureOnly = FALSE,  
                                      removeSubjectsWithPriorOutcome = TRUE,  
                                      priorOutcomeLookback = 9999,  
                                      riskWindowStart = 1,  
                                      risk  
  
##  
  
###
```

viewPlp

R

runPLp

ATLAS

```
plpResult <- loadPlpResult(file.path(outputFolder,  
                                     'Analysis_1',  
                                     'plpResult'))
```

Analysis_1

Shiny

```
viewPlp(plpResult)
```

Shiny

13.10 AUC 0.78 0.74

13.11

13.12

13.13 40

attrition

13.14

13.7.7

ATLAS

The screenshot shows a Shiny web application interface. At the top, there's a header bar with a logo, the URL "http://127.0.0.1:4888", an "Open in Browser" button, and a "Publish" dropdown. Below the header, the title "PatientLevelPrediction Explorer" is followed by tabs for "Internal Validation" and "External Validation". A navigation bar below the tabs includes "Evaluation Summary" (which is active, indicated by a blue border), "Characterization", "ROC", "Calibration", "Demographics", "Preference", "Box Plot", and "Settings".

The main content area is titled "Evaluation Summary". It features a table with 11 rows of data. The table has three columns: "Metric" (leftmost), "test" (middle), and "train" (rightmost). The "Metric" column lists various performance metrics. The "test" and "train" columns show their respective values.

Metric	test	train
1 AUC	0.72130	0.75348
2 AUC_lb95ci	0.70057	0.74215
3 AUC_ub95ci	0.74203	0.76482
4 AUPRC	0.10971	0.13571
5 BrierScaled	0.03755	0.04902
6 BrierScore	0.03355	0.03304
7 CalibrationIntercept.Intercept	-0.00089	-0.00813
8 CalibrationSlope.Gradient	1.02041	1.22457
9 outcomeCount	601.00000	1802.00000
10 populationSize	16685.00000	50054.00000
11 Incidence	3.60204	3.60011

At the bottom left, it says "Showing 1 to 11 of 11 entries". On the right, there are buttons for "Previous", a page number "1", and "Next".

Figure 13.10: Shiny

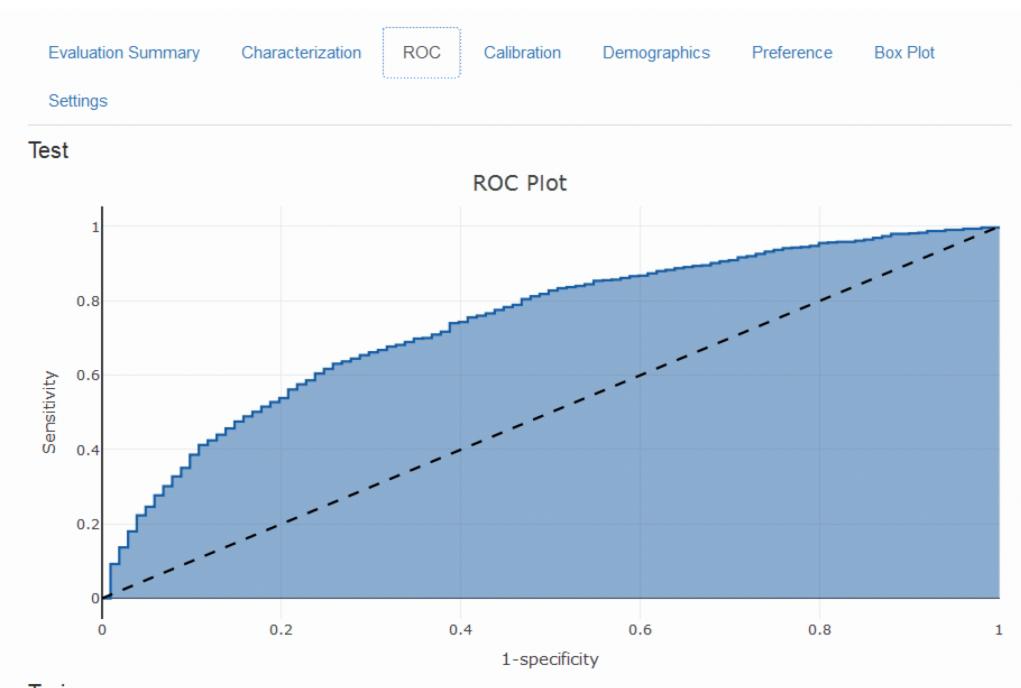


Figure 13.11: ROC Plot

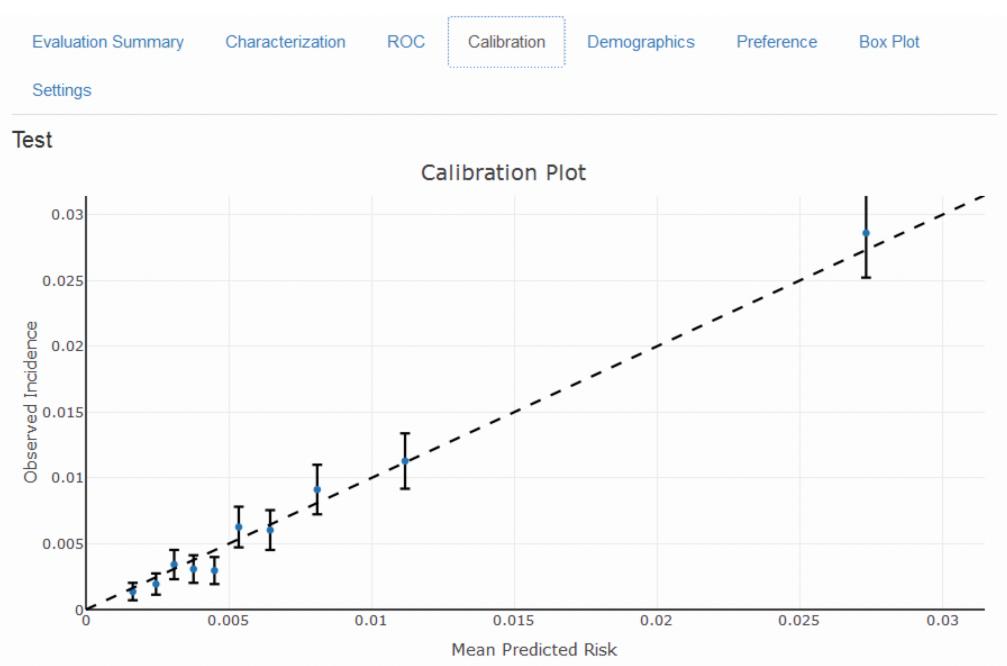


Figure 13.12: Calibration Plot

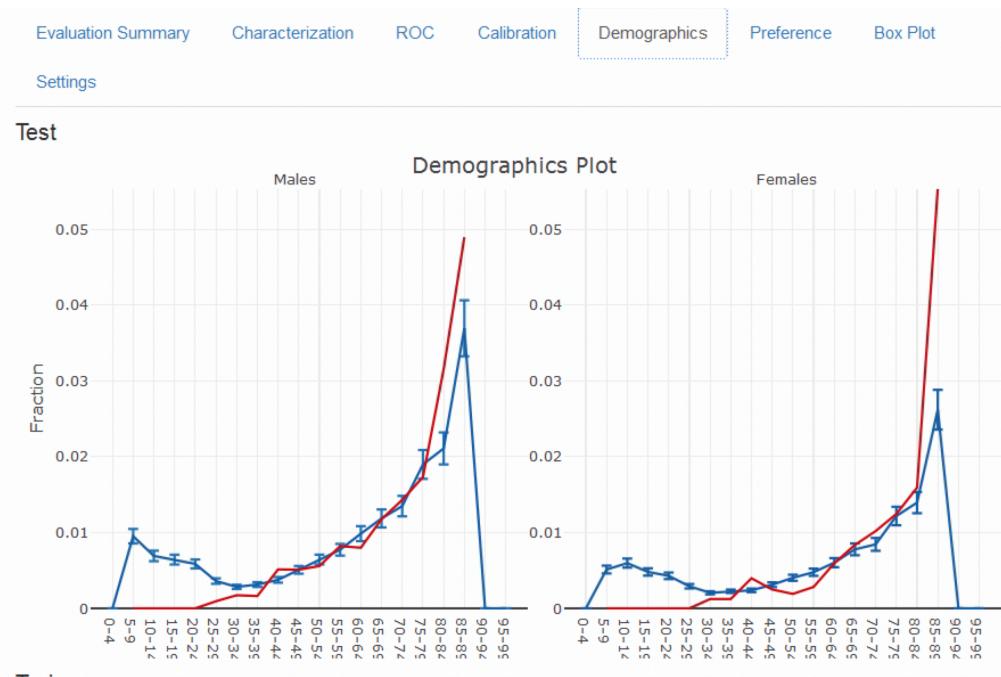


Figure 13.13: □□□□□□□□□□□□□□□□□□

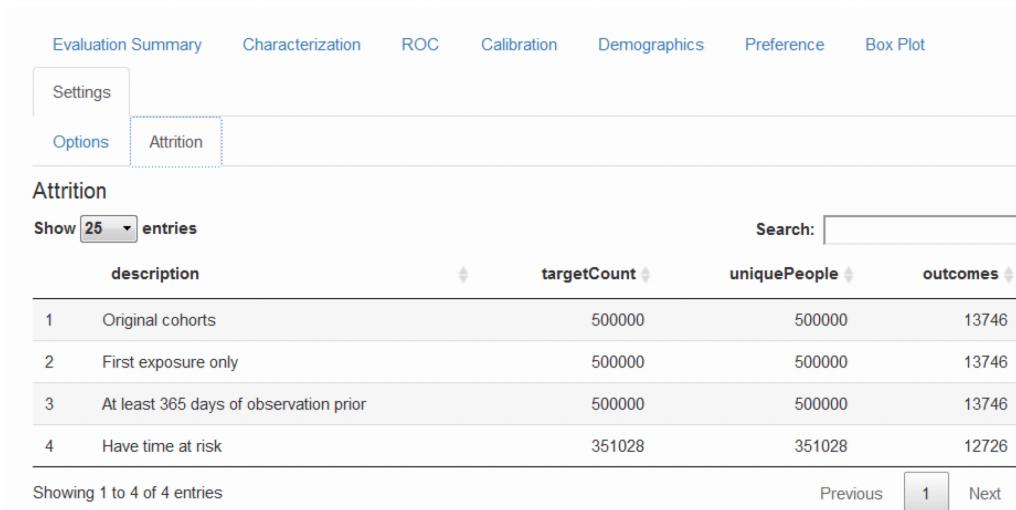


Figure 13.14: attrition

Shiny 13.15

Figure 13.15: Shiny

A horizontal row of 20 empty square boxes, intended for children to practice writing their names.

- AUC
 - AUPRC

A horizontal row of 20 empty square boxes for writing responses.

Settings

Setting	Value
1	Model
2	variance
3	seed

Figure 13.16: □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□

Performance 13.17

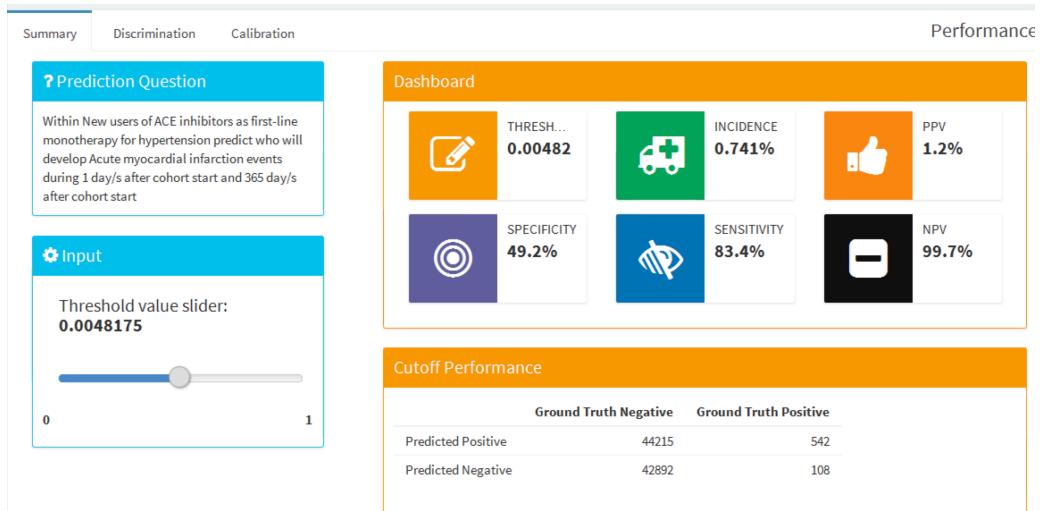


Figure 13.17: □□□□□□□□□□□□□□□□□□□□□□□□□□□□□

Discrimination ROC -
ROC - ROC

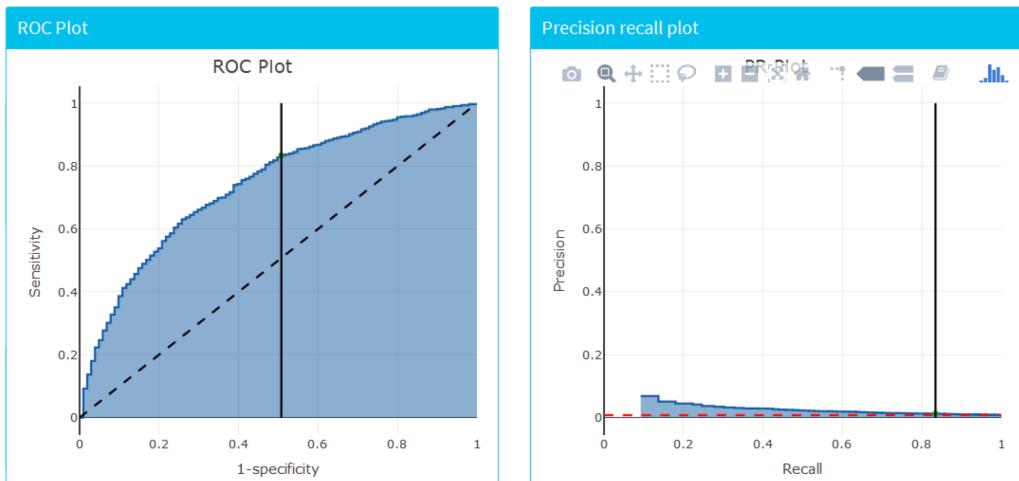


Figure 13.18: ROC curve - Confusion matrix

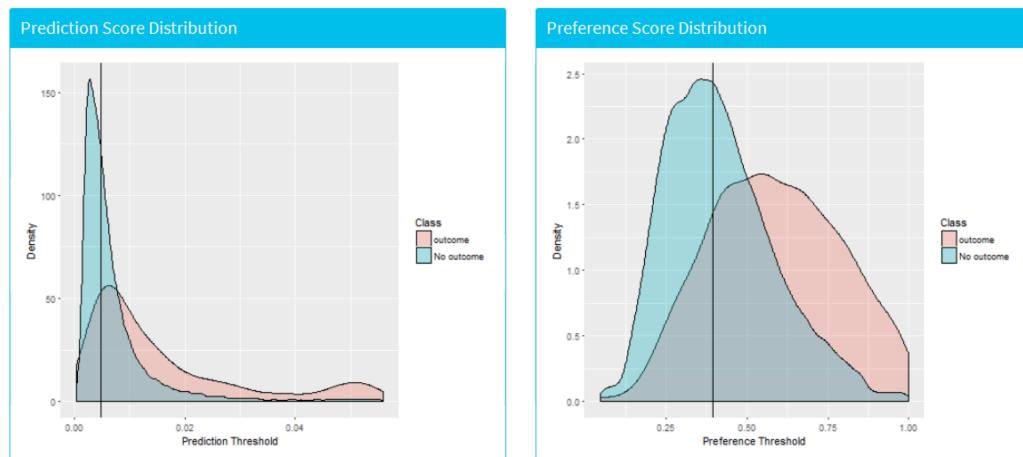


Figure 13.19: Prediction Score Distribution and Preference Score Distribution plots.

13.20 Calibration Plot and Demographic Plot

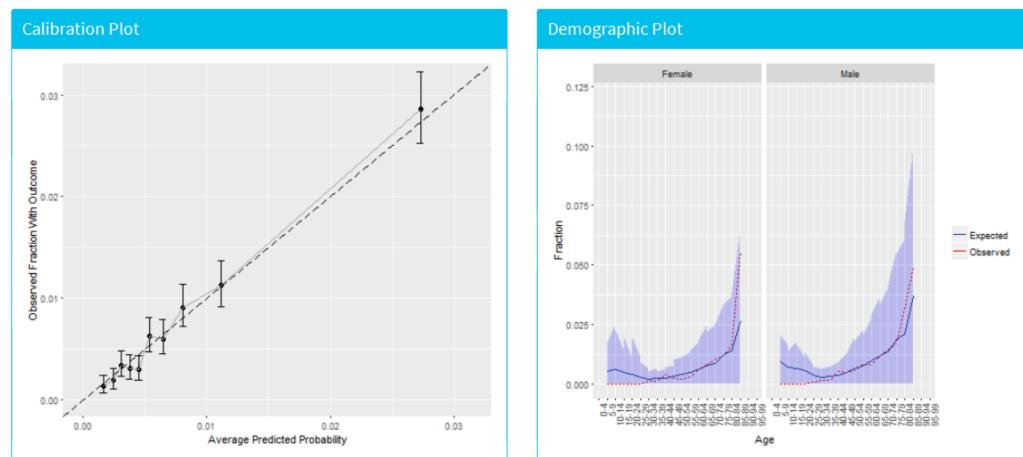


Figure 13.20: Calibration Plot and Demographic Plot.

13.21 Model



13.21 Model

13.22

13.22

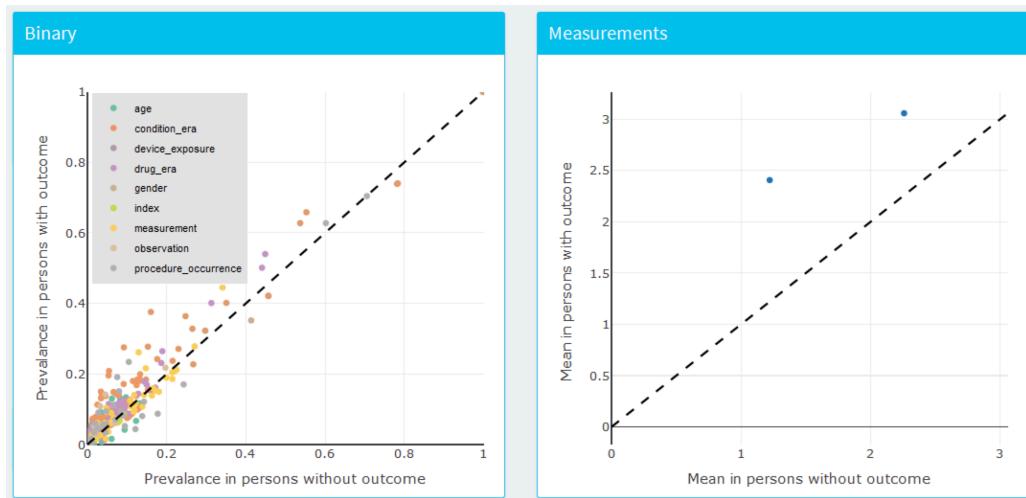


Figure 13.21: □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□

	Covariate Name	Value	Outcome Mean	Non-outcome Mean
1	age group: 00-04	0	0.0004	0.0001
2	age group: 05-09	0	0	0.0003
3	index month: 1	0	0.1307	0.1096
4	observation during day -365 through 0 days relative to index: Domain	0	0.1188	0.0514
5	Charlson index - Romano adaptation	0	2.4783	1.3817
6	Diabetes Comorbidity Severity Index (DCSI)	0.1478	2.4056	1.2207
7	CHADS2VASc	0.9279	3.0573	2.2576
8	visit_occurrence concept count during day -365 through 0 concept_count relative to index	0	19.5263	13.8837
9	age group: 10-14	0	0	0.001
10	index month: 2	0	0.0934	0.0909

Figure 13.22: □□□□□□□□□□



13.22

13.8

13.8.1

```
createPlpJournalDocument(plpResult = <your plp results>,
                         plpValidation = <your validation results>,
                         plpData = <your plp data>,
                         targetName = "<target population>",
                         outcomeName = "<outcome>",
                         table1 = F,
                         connectionDetails = NULL,
                         includeTrain = FALSE,
                         includeTest = TRUE,
                         includePredictionPicture = TRUE,
                         includeAttritionPlot = TRUE,
                         outputLocation = "<your location>")
```

13.9



13.10

100

84.5 R R-Studio Java

```
install.packages(c("SqlRender", "DatabaseConnector", "remotes"))
remotes::install_github("ohdsi/Eunomia", ref = "v1.0.0")
remotes::install_github("ohdsi/PatientLevelPrediction")
```

Eunomia CDM R

```
connectionDetails <- Eunomia::getEunomiaConnectionDetails()
```

```
Eunomia::createCohorts(connectionDetails)
```

1 / 4

NSAIDs GI

NSAID COHORT_DEFINITION_ID 4 GI COHORT_D

E.9

Part II

— — —

Chapter 14

□□□□□□□: *Patrick Ryan & Jon Duke*

14.1

Desired attribute	Question	Researcher	Data	Analysis	Result
Repeatable	Identical	Identical	Identical	Identical	= Identical
Reproducible	Identical	Different	Identical	Identical	= Identical
Replicable	Identical	Same or different	Similar	Identical	= Similar
Generalizable	Identical	Same or different	Different	Identical	= Similar
Robust	Identical	Same or different	Same or different	Different	= Similar
Calibrated	Similar (controls)	Identical	Identical	Identical	= Statistically consistent

Figure 14.1: □□□□□□□□□□□□□□□□

Austin Bradford Hill (1965) et al., 2013a) et al., 2016, 2018a,b)

14.2

(Bots et al., 2010; Hersh et al., 2013; Sherman et al., 2016) et al., 2012; Liaw et al., 2013; Weiskopf and Weng, 2013) OHDSI CDM OHDSI (Huser et al., 2016; Kahn et al., 2015; Callahan et al., 2017; Yoon et al., 2016)

A horizontal row of 30 small, empty rectangular boxes, likely used for input fields or placeholder text in a form.

□ : □ □ □ □ □ 4 □ □ □ □ □ □ □ □ □ □

A horizontal sequence of 20 empty square boxes, intended for students to draw their own shapes or patterns.

14.3

##



Chapter 15

□□□□□: Martijn Schuemie, Vojtech Huser & Clair Blacketer

A horizontal row of 20 empty square boxes, intended for children to write their names in, likely as part of a classroom activity or name recognition exercise.

□□□□□ DQ□□□□□□□□□□□ (Roebuck, 2012):

□□□□□□□ DQ □□□□□□□ DQ □□□□□□□

15.1

Chapter 14 and Johnson (2003) DQ

15.1.2 DQ (Defalco et al., 2013; Makadia and Ryan, 2014; Matcho et al., 2014; Voss et al., 2015a,b; Hripcak et al., 2018) CDM

3 OHDSI DQ
16 Chapter 17 Chapter 18 #

Kahn et al.
(2016) DQ 3

□ □ □ □ □ □ □ □ □ **2** □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □

15.1.1

ACHILLES (Huser et al., 2018) ACHILLES (Huser et al., 2016) ACHILLES ATLAS

ATLAS

(DQD) ACHILLES CDM CDM
1,500 FAIL 15.1.1

□: □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□

0.34	VISIT_OCCURRENCE	<input checked="" type="checkbox"/> provider_id	0.05	FAIL
		<input type="checkbox"/> visit_occurrence_id		

			□ □	□ □
0.99	MEASUREMENT □□□□□ measurement_source_value □□□□□□□□□□□□□□□□□□□□□□□□□ 0 □□□□□□□□□□□	0.30		FAIL
0.09	DRUG_ERA □□□□□ drug_concept_id □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□	0.10		PASS
0.02	DRUG_EXPOSURE □□□□□ DRUG_EXPOSURE_END_DATE □□□□□□□ DRUG_EXPOSURE_START_DATE □□□□□□□□□□□□□□□□□□□□□□□□□	0.05		PASS
0.00	PROCEDURE_OCCURRENCE □□□□□ procedure_occurrence_id □□□□□□□□□□□□□□□□□□□□□□□□□	0.00		PASS

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CDM □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□
CDM □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□
ID □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□
ID □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□



ACHILLES □ DQD □ CDM □□□□□□□□□□□□□□□□□□
 CDM □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□

15.1.2 ETL

□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□
CDM □□□□□ ETL (Extract-Transform-Load) □□□□□□□□□□□□□□
 □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□ ETL
 ETL □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□
 6 □□□□□ Rabbit-in-a-Hat □□□□□□□□□□□□□□□□□□
 □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□ CDM □□□□□□□□□□ R □□□□□□□□□□□□□□□□
 CDM □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□

```
source("Framework.R")
declareTest(101, "Person gender mappings")
add_enrollment(member_id = "M000000102", gender_of_member = "male")
add_enrollment(member_id = "M000000103", gender_of_member = "female")
expect_person(PERSON_ID = 102, GENDER_CONCEPT_ID = 8507)
expect_person(PERSON_ID = 103, GENDER_CONCEPT_ID = 8532)
```

Rabbit-in-a-Hat gender mappings ENROLLMENT
Rabbit-in-a-Hat add_enrollment MEMBER_ID
GENDER_OF_MEMBER 2 ETL
PERSON 2

ENROLLMENT White Rabbit

ETL SQL ETL SQL

```
insertSql <- generateInsertSql(databaseSchema = "source_schema")
testSql <- generateTestSql(databaseSchema = "cdm_test_schema")
```

□□□□□□□□□ 15.1 □□□□□□□□□

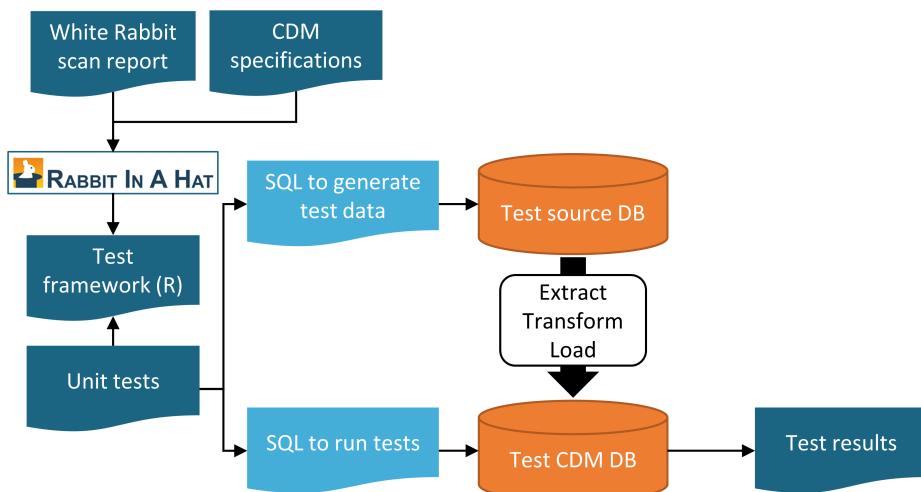


Figure 15.1: Rabbit-in-a-Hat ☐☐☐☐☐☐☐☐☐☐☐☐☐☐ ETL (Extract-Transform-Load) ☐☐☐☐☐☐☐☐☐☐☐

□: ETL □□□□□□□□□□

ID	□ □	□ □
101	Person gender mappings	PASS
101	Person gender mappings	PASS

ETL

15.2

15.2.1

Vocabulary
Vocabulary¹
MethodEvaluation R
checkCohortSourceCodes ATLAS
15.2
440383 (“ ”) 3
9 3.11 ICD-10 F32.8 F32.89
9 ICD-10
10 ICD-9 ICD-
10 F32.9

% per month	Max monthly %	Person count	Description
	26.81	92,019,885	Depressive Disorder
	6.64	15,969,198	Depressive disorder 440383
	6.64	15,686,275	311 (ICD9CM) Depressive disorder, not elsewhere classified
	0.46	188,230	F328 (ICD10CM) Other depressive episodes
	0.38	94,693	F3289 (ICD10CM) Other specified depressive episodes
	3.10	12,010,783	Adjustment disorder with mixed emotional features 433454
	3.07	9,839,712	30928 (ICD9CM) Adjustment disorder with mixed anxiety and depressed mood
	3.03	2,049,618	F4323 (ICD10CM) Adjustment disorder with mixed anxiety and depressed mood
	0.04	121,453	3091 (ICD9CM) Prolonged depressive reaction
	3.17	9,237,192	Dysthymia 433440

Figure 15.2: checkCohortSourceCodes

R **findOrphanSourceCodes** ICD-10 J85.0 4324261

¹<https://github.com/OHDSI/Vocabulary-v5.0/issues>

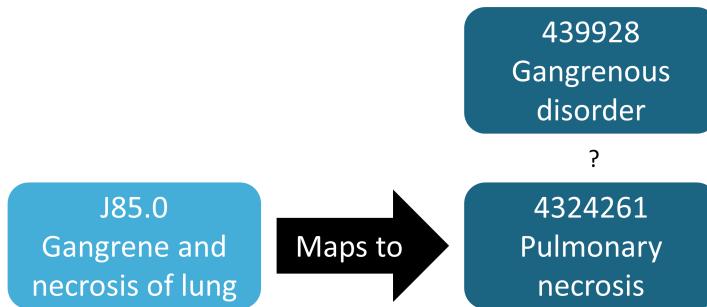


Figure 15.3: □□□□□□□□□□□□□□□□

15.3 ACHILLES

□□□□□CDM□□□□□□□□□□□□□□□ACHILLES□□□□□□□□□□□

```
library(Achilles)
connDetails <- createConnectionDetails(dbms = "postgresql",
                                         server = "localhost/ohdsi",
                                         user = "joe",
                                         password = "supersecret")

cdmDbSchema <- "my_cdm_data"
cdmVersion <- "5.3.0"
```

ACHILLES

```
result <- achilles(connectionDetails,  
                     c  
## Data Quality Dashboard  {#dqInPractice}
```

CDM Data Quality Dashboard \c@ref(achillesInPractice) CDM

```
cdmDbSchema <- "my_cdm_data.dbo"
```

Dashboard

```
DataQualityDashboard::executeDqChecks(connectionDetails = connectionDetails,
                                         cdmDatabaseSchema = cdmDbSchema,
                                         resultsDatabaseSchema = cdmDbSchema,
                                         cdmSourceName = "My database",
                                         outputFolder = "My output")
```

viewDqDashboard(jsonPath)

(jsonPath) Dashboard JSON
Dashboard 15.4 Kahn



Figure 15.4: Data Quality Dashboard

(Results) 15.5

15.4

B.4

```
library(MethodEvaluation)
json <- readChar("cohort.json", file.info("cohort.json")$size)
sql <- readChar("cohort.sql", file.info("cohort.sql")$size)
checkCohortSourceCodes(connectionDetails,
```

RESULTS

SYNTHIA SYNTHETIC HEALTH DATABASE

Results generated at 2019-08-22 14:15:06 in 29 mins

							Column visibility	CSV
						Search:		
STATUS	CONTEXT	CATEGORY	SUBCATEGORY	LEVEL	DESCRIPTION	% RECORDS		
				FIELD				
+	FAIL	Verification	Plausibility	Atemporal	FIELD The number and percent of records with a value in the gap_days field of the DRUG_ERA table less than 0. (Threshold=0%).	24.07%		
+	FAIL	Verification	Completeness	None	FIELD The number and percent of records with a value of 0 in the standard concept field race_concept_id in the PERSON table. (Threshold=0%).	16.74%		
+	FAIL	Verification	Conformance	Relational	FIELD The number and percent of records that have a value in the ethnicity_concept_id field in the PERSON table that does not exist in the CONCEPT table. (Threshold=0%).	16.15%		
+	PASS	Verification	Completeness	None	FIELD The number and percent of records with a NULL value in the condition_end_date of the CONDITION_OCCURRENCE. (Threshold=100%).	13.24%		
+	PASS	Verification	Completeness	None	FIELD The number and percent of records with a NULL value in the condition_end_datetime of the CONDITION_OCCURRENCE. (Threshold=100%).	13.24%		

Showing 71 to 75 of 1,327 entries (filtered from 1,639 total entries) Previous 1 ... 14 15 16 ... 266 Next

Figure 15.5: Data Quality Dashboard

```
cdmDatabaseSchema = cdmDbSchema,
cohortJson = json,
cohortSql = sql,
outputFile = "output.html")
```

15.6

9 ICD-10

Angioedema

```
orphans <- findOrphanSourceCodes(connectionDetails,
                                    cdmDatabaseSchema = cdmDbSchema,
                                    conceptName = "Angioedema",
                                    conceptSynonyms = c("Angioneurotic edema",
                                                       "Giant hives",
                                                       "Giant urticaria",
                                                       "Periodic edema"))
```

View(orphans)

□□□

□□

□□ID

□□□□□□

T78.3XXS Angioneurotic edema, sequela
10002425 Angioedemas

ICD10CM 508
MedDRA 0

□□□	□□	□□ID	□□□□□□□□
148774	Angioneurotic Edema of Larynx	CIEL	0
402383003	Idiopathic urticaria and/or angioedema	SNOMED	0
232437009	Angioneurotic edema of larynx	SNOMED	0
10002472	Angioneurotic edema, not elsewhere classified	MedDRA	0

“Angioneurotic edema, sequelae”

15.5



15.6

10 of 10

8.4.5 R-R-Studio Java

```
install.packages(c("SqlRender", "DatabaseConnector", "remotes"))
remotes::install_github("ohdsi/Achilles")
remotes::install_github("ohdsi/DataQualityDashboard")
remotes::install_github("ohdsi/Eunomia", ref = "v1.0.0")
```

Eunomia R CDM

```
connectionDetails <- Eunomia:::getEunomiaConnectionDetails()
```

CDM main

15.1. Eunomia ACHILLES

15.2. Eunomia Data Quality Dashboard

15.3. DOD

% per month	Max monthly %	Person count	Description
	60.60	24,189,656	Inpatient or ER visit
	39.50	15,003,249	Emergency Room Visit 9203
	39.50	15,003,249	ER (None) No matching concept
	23.90	9,186,407	Inpatient Visit 9201
	23.90	9,186,407	IP (None) No matching concept
	0.27	76,711	Angioedema
	0.27	76,711	Angioedema 432791
	0.26	64,726	9951 (ICD9CM) Angioneurotic edema, not elsewhere classified
	0.20	8,822	T783XXA (ICD10CM) Angioneurotic edema, initial encounter
	0.09	3,163	T783XXD (ICD10CM) Angioneurotic edema, subsequent encounter

Figure 15.6: □□□

Chapter 16

□□□□□□□: Joel Swerdel, Seng Chan You, Ray Chen & Patrick Ryan

OHDSI

16.1

```
ACE ACE  
## {#CohortValidation}
```

16.1.1

		Gold Standard	
		True	False
Cohort Definition	True	True Positive	False Positive
	False	False Negative	True Negative

Figure 16.1: □□□□□□□□□□□□□□

incorrect

1. **PPV** = $\frac{\text{True Positives}}{\text{True Positives} + \text{False Positives}}$
 2. **NPV** = $\frac{\text{True Negatives}}{\text{True Negatives} + \text{False Negatives}}$
 3. **PPV** = $\frac{\text{True Positives}}{\text{True Positives} + \text{False Positives}}$
 4. **NPV** = $\frac{\text{True Negatives}}{\text{True Negatives} + \text{False Negatives}}$

100% Rubbo et al. (2015) 33 1

PPV 1

16.1.2

1. OHDSI IRB
 2. ATHENA ATLAS
4163261 ID
4314337 ID 4048666
 3. ATLAS PERS
 4. EHR
 5. 1
 - 6.
 7. 1 2 3
 8. PPV
 9. CUIMC
 10. PPV

MI Rubbo et al. (2015) # PheEvaluator

OHDSI et al., 2019) PheValuator Patient Level Prediction

Patient-Level Prediction Chapter 13 (et al., 2013)

16.1.3 PheEvaluator

MI □

¹<https://github.com/OHDSI/PheEvaluator>

□ □ □ □ 1: xSpec □ □ □ □ □ □

16.2 ATLAS MI xSpec

Cohort #10934

MI xSpec Cohort

Definition Concept Sets Generation Reporting Export

[460] MI xSpec Model

Cohort Entry Events

Events having any of the following criteria:

a condition occurrence of [460] Myocardial Infarction

+ Add attribute...

with continuous observation of at least days before and days after event index date

Limit initial events to: earliest event per person.

Restrict initial events to:

having all of the following criteria:

+ Add criteria to group...

with at least days using all occurrences of:

a condition occurrence of [460] Myocardial Infarction

+ Add attribute...

with a Visit occurrence of:

where event starts between days Before and days After index start date [add additional constraint](#)

restrict to the same visit occurrence

allow events from outside observation period

and with at least days using all occurrences of:

a condition occurrence of [460] Myocardial Infarction

+ Add attribute...

where event starts between days After and days After index start date [add additional constraint](#)

restrict to the same visit occurrence

allow events from outside observation period

Limit initial events to: earliest event per person.

Remove initial event restriction

Figure 16.2: xSpec

□ □ □ □ 2: xSens □ □ □ □ □ □

16.3 ATLAS MI xSens

□ □ □ □ 3: □ □ □ □ □ □ □ □

```
createPhenoModel
```

The screenshot shows the 'MI xSens Cohort' page. The top navigation bar includes a logo, the cohort name, and several icons for file operations. Below the header is a menu bar with tabs: 'Definition' (selected), 'Concept Sets', 'Generation', 'Reporting', and 'Export'. A large text input field below the menu is labeled 'enter a cohort definition description here'. The main content area is titled 'Cohort Entry Events' and contains instructions for defining events. It features a dropdown menu for selecting event types, a button to add initial events, and a button to delete criteria. Below this, there are fields for specifying observation periods and limiting initial events per person. A green button at the bottom left allows users to restrict initial events.

MI xSens Cohort

Definition Concept Sets Generation Reporting Export

enter a cohort definition description here

Cohort Entry Events

Events having any of the following criteria:

a condition occurrence of [460] Myocardial Infarction ▾

+ Add Initial Event ▾

+ Add attribute... ▾

Delete Criteria

with continuous observation of at least 0 days before and 0 days after event index date

Limit initial events to: earliest event ▾ per person.

Restrict initial events

Figure 16.3: xSens

```
setwd("c:/temp")
library(PheEvaluator)
connectionDetails <- createConnectionDetails(
  dbms = "postgresql",
  server = "localhost/ohdsi",
  user = "joe",
  password = "supersecret")

phenoTest <- createPhenoModel(
  connectionDetails = connectionDetails,
  xSpecCohort = 10934,
  cdmDatabaseSchema = "my_cdm_data",
  cohortDatabaseSchema = "my_results",
  cohortDatabaseTable = "cohort",
  outDatabaseSchema = "scratch.dbo", #
  trainOutFile = "5XMI_train",
  exclCohort = 1770120, #xSens
  prevCohort = 1770119, #
  modelAnalysisId = "20181206V1",
  excludedConcepts = c(312327, 314666),
  addDescendantsToExclude = TRUE,
  cdmShortName = "myCDM",
  mainPopnCohort = 0, #
  lowerAgeLimit = 18,
```

```
upperAgeLimit = 90,  
gender = c(8507, 8532),  
startDate = "20100101",  
endDate = "20171231")
```

createEvalCohort □□□□ PatientLevelPrediction □□□□□ applyModel □□□□□□□□□□□□□

□□□□

```
setwd("c:/temp")
connectionDetails <- createConnectionDetails(
  dbms = "postgresql",
  server = "localhost/ohdsi",
  user = "joe",
  password = "supersecret")

evalCohort <- createEvalCohort(
  connectionDetails = connectionDetails,
  xSpecCohort = 10934,
  cdmDatabaseSchema = "my_cdm_data",
  cohortDatabaseSchema = "my_results",
  cohortDatabaseTable = "cohort",
  outDatabaseSchema = "scratch.dbo",
  testOutFile = "5XMI_eval",
  trainOutFile = "5XMI_train",
  modelAnalysisId = "20181206V1",
  evalAnalysisId = "20181206V1",
  cdmShortName = "myCDM",
  mainPopnCohort = 0,
  lowerAgeLimit = 18,
  upperAgeLimit = 90,
  gender = c(8507, 8532),
  startDate = "20100101",
  endDate = "20171231")
```

{#GeneralizabilityOfEvidence}

OHDSI vs. **Other**
vs. **Other** vs. **Other**
vs. **Other**

16.2



Chapter 17

— — — — —

□□□□□□□: *Martijn Schuemie*

A horizontal row of 20 empty square boxes, intended for students to draw or write in.

17.1.1 CDM

8.1. OHDSI

Methods Library

Methods Library

17.1

17.1.1

OHDSI □ R □ DatabaseCo

CRM OHDSI

Methods Library □□□ B □□□□□□□□□□□□□□□□□□

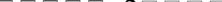
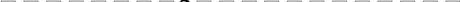
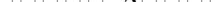
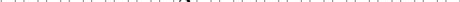
Michigan Library _____

17.1.2

2008) — 4

- `sampledPatients`: `sampledPatients <- takeSample(patients, sampleSize = 100)`
 - `1`: `1 <- sampledPatients[1]`

17.1.3

-  : 1  1 
 -  : 2  2 

17.1.4 Methods Library

OHDSI Methods Library

GitHub drat

OHDSI License V2 R C++ SQL Java OHDSI

17.1.5

GitHub git

OHDSI

- <img

¹<https://github.com/OHDSI/CohortMethod/issues>

²<http://forums.ohdsi.org/>

17.1.6

R 1

17.1.7

GitHub drat

17.1.8

OHDSI

17.1.9

OHDSI

17.1.10

OHDSI GitHub⁴ GitHub https://github.com/security OHDSI

17.1.11

OHDSI GitHub GitHub https://github.com/security

17.2

17.2.1

17.2.2

³<https://ohdsi.github.io/MethodsLibrary/>

⁴<https://github.com/>

17.3



Chapter 18

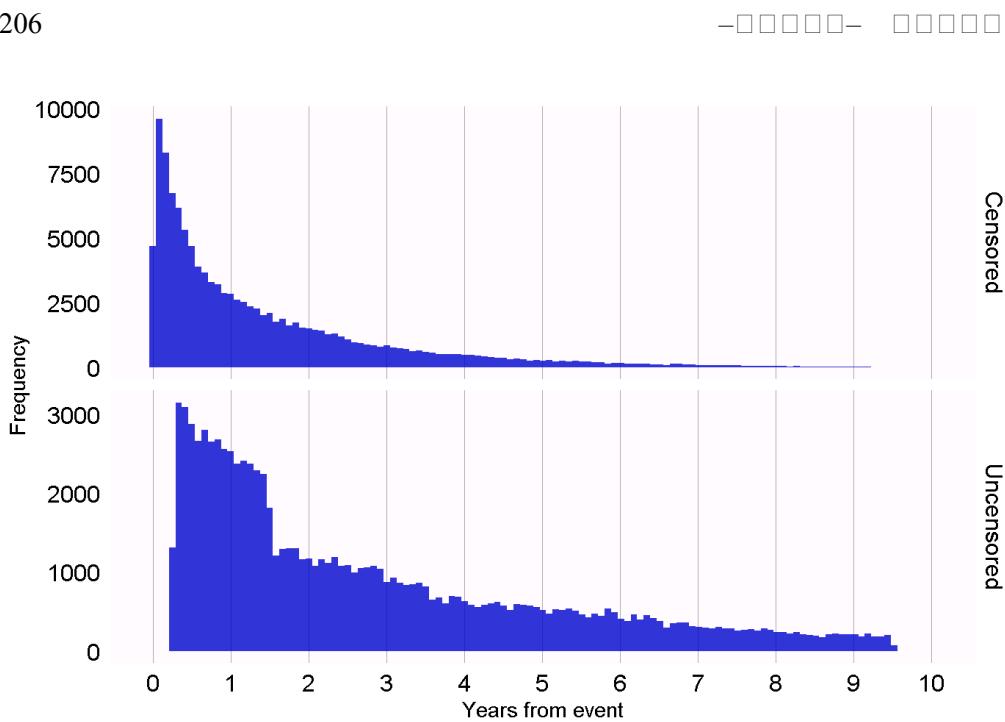
□□□□□: Martijn Schuemie

A horizontal row of 30 empty square boxes, intended for children to write their names in, likely as part of a classroom activity or name recognition exercise.

18.1

OHDSI #ref(studyOutputs) CohortMethod

18.2



18.2.1

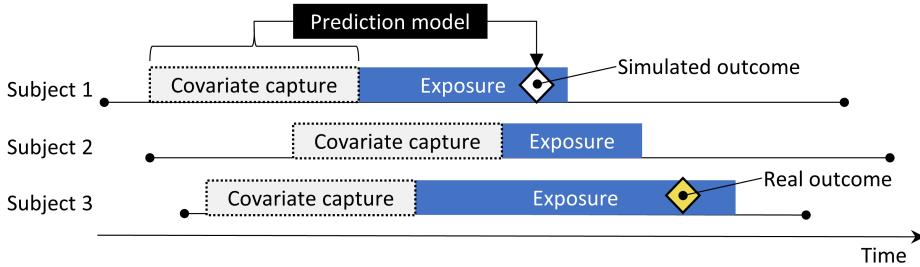
et al., 2016) et al., 2016)

18.2.2

et al., 2014)

OHDSI

et al., 2013) 10



18.2.3

- **AUC**
 - 95
 - $1/()^2$
 - **MSE**
 - $1 \alpha = 0.05$
 α
 - $2 \alpha = 0.05$
 $1 - \alpha$
 -

18.2.4 P

$\alpha = 0.05$ (et al., 2014)

$\hat{\theta}_i$ τ_i $i = 1, \dots, n$
 $\theta_i = 0$ i β_i θ_i
 β_i $p(\hat{\theta}_i | \theta_i + \beta_i)$
 $\hat{\tau}_i^2 p(\beta_i | \beta_i)$
 $\beta_i \sim \mu + \sigma^2 \mu$
 $\sigma^2 \mu$

A horizontal row of fifteen empty square boxes, intended for children to write their names in, likely as part of a classroom activity or name recognition exercise.

$$\beta_i \sim N(\mu, \sigma^2) \quad \square \square \square \quad \hat{\theta}_i \sim N(\theta_i + \beta_i, \tau_i^2)$$

$$L(\mu, \sigma | \theta, \tau) \propto \prod_{i=1}^n \int p(\hat{\theta}_i | \beta_i, \theta_i, \hat{\tau}_i) p(\beta_i | \mu, \sigma) d\beta_i$$

$\hat{\mu}$ $\hat{\sigma}$ $\hat{\theta}_{n+1}$ $\hat{\tau}_{n+1}$ {n##}

ACEi THZ

18.2.5

100

ACEi and THZ combined									Optimize								
Concept Set Expression		Included Concepts 14		Included Source Codes		Explore Evidence		Export		Compare							
Show 25 ▾ entries					Search: <input type="text"/>					Previous 1	Next						
Showing 1 to 14 of 14 entries																	
Concept Id	Concept Code	Concept Name	Domain	Standard Concept Caption	<input type="checkbox"/> Exclude	<input type="checkbox"/> Descendants	<input type="checkbox"/> Mapped										
1342439	38454	trandolapril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>										
1334456	35296	Ramipril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>										
1331235	35208	quinapril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>										
1373225	54552	Perindopril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>										
1310756	30131	moexipril	Drug	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>										

C.1 76

Evidence for all conditions for ACEi and THZ combined

18.2.6

Query Language Chapter
SQL R SQL R

OHDSI

```
library(MethodEvaluation)
```

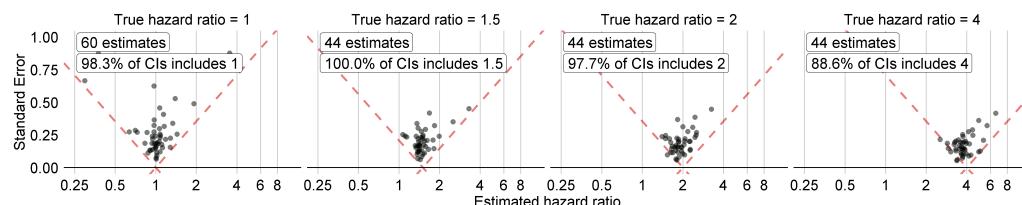
```
    outcomeId = ncs)

pcs <- synthesizePositiveControls(
  connectionDetails = connectionDetails,
  cdmDatabaseSchema = cdmDbSchema,
  exposureDatabaseSchema = cohortDbSchema,
  exposureTable = cohortTable,
  outcomeDatabaseSchema = cohortDbSchema,
  outcomeTable = cohortTable,
  outputDatabaseSchema = cohortDbSchema,
  outputTable = cohortTable,
  createOutputTable = FALSE,
  modelType = "survival",
  firstExposureOnly = TRUE,
  firstOutcomeOnly = TRUE,
```

```
removePeopleWithPriorOutcomes = TRUE,  
washoutPeriod = 365,  
riskWindowStart = 1,  
riskWindowEnd = 0,  
endAnchor = "cohort end",  
exposureOutcomePairs = eoPairs,  
effectSizes = c(1.5, 2, 4),  
cdmVersion = cdmVersion,  
workFolder = file.path(outputFolder, "pcSynthesis"))
```

Methods Library

18.2.7



[MethodEvaluation](<https://ohd ## OHDSI>)

18 3

¹<http://data.ohdsi.org/MethodEvalViewer/>



Figure 18.6: 95% GI = IBD =

Part III

OHDSI

Chapter 19

□□□□□□□□: *Sara Dempster & Martijn Schuemie*

OHDSI OHDSI OHDSI R SQL CDM OMOP
ETL 6 #

19.0.1

Level Estimation 12 / 18

19.0.2

Level Prediction PLE EHR

19.0.3

19.0.4

OHDSI 2, 7, 11, 12, 13

OHDSI ??

19.0.5

19.0.6 CDM

OHDSI OMOP CDM
4 ETL
CDM

CDM OMOP 5 OHDSI
CDM OMOP
#

19.0.7

(T2DM) Chapter 7 3
2 2
T2DM T2DM
T2DM T2DM
OHDSI
7

19.0.8

Chapter 15 T2DM T2DM T2DM PLE

1909

OHDSI ATLAS
@ref(Ohd ##)



Chapter 20

— — OHDSI

□□□□: Kristin Kostka, Greg Klebanov & Sara Dempster

OHDSI 19 Research Network

20.1 OHDSI

OHDSI CDM



□□□□□ OHDSI □□□□□□□□□□□□

- OHDSI
 - OMOP
 - OMOP-CDM

20.2 OHDSI

20.2.1 OHDSI

20.2.2 OHDSI



OHDSI CDM

OHDSI OMOP CDM

OHDSI OHDSI OHDSI CDM OHDSI

OHDSI OHDSI OHDSI

- OHDSI GitHub
- CDM R SQL
- OHDSI OHDSI
- OHDSI GitHub
- R Shiny data.ohdsi.org

OHDSI

20.2.3 OHDSI

OHDSI

OHDSI

OHDSI

20.2.4 OHDSI

OHDSI OHDSI OHDSI CDM OHDSI OHDSI

OHDSI

- OHDSI
- CDM RStudio
- OHDSI R
- OHDSI R
- OHDSI R



Chapter ?(ExtractTransformLoad) ETL
?(DataQuality)

Level Effect Estimation PLE Patient Level Prediction PLP OHDSI

Level Effect Estimation PLE Patient Level Prediction PLP OHDSI

OHDSI

OHDSI

- OHDSI
- OHDSI
- OHDSI

20.2.5

CDM CMS Synthetic Public
Use Files Mitre SyntheticMass Synthea OHDSI
19 IRB OHDSI

20.2.6

OHDSI OHDSI OHDSI
GitHub OHDSI OHDSI OHDSI
SFTP Amazon S3
/ OHDSI
OHDSI
OHDSI

20.2.7

OHDSI GitHub Shiny



OHDSI JANE Journal/Author
Name Estimator¹

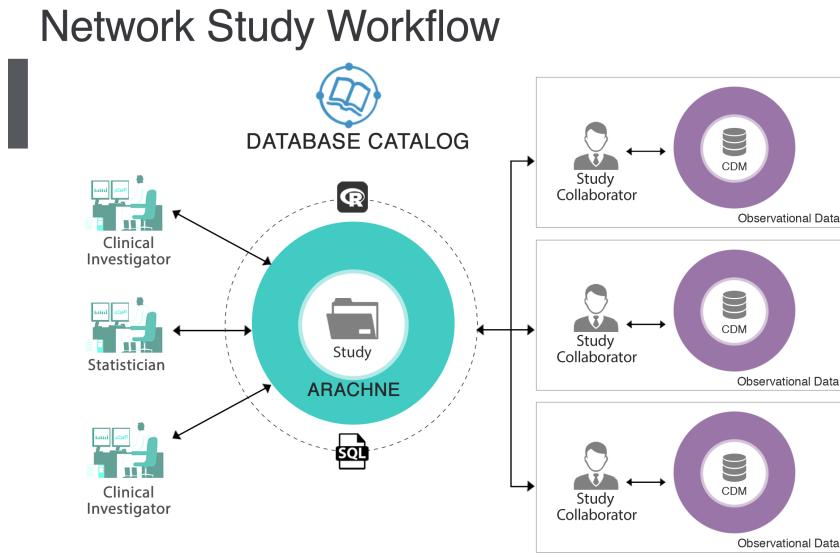


Figure 20.1: ARACHNE

ARACHNE ACHILLES ATLAS

20.3 OHDSI

OHDSI OHDSI

GitHub <https://github.com/ohdsi> data.ohdsi.org R Shiny CDM OHDSI 4 CDM

²<http://www.icmje.org/recommendations/browse/roles-and-responsibilities/defining-the-role-of-authors-and-contributors.html> ## □□: □□□□□□□□□□□□□□□□□□

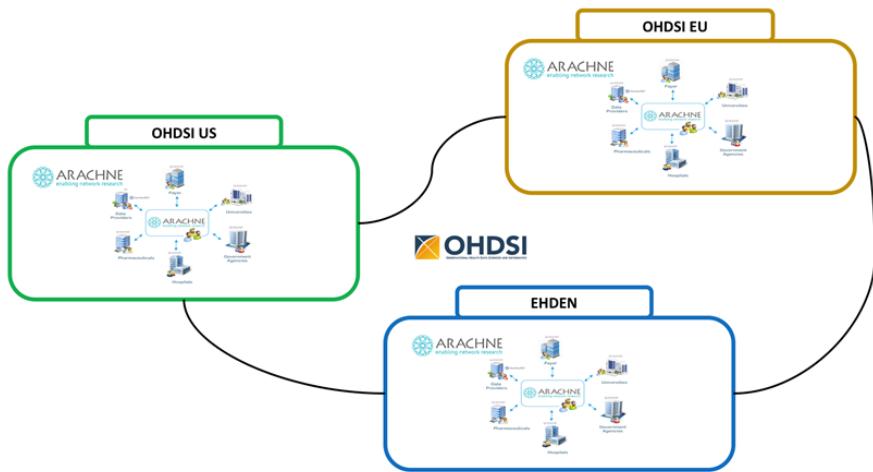


Figure 20.2: ARACHNE

20.4



- OHDSI CDM OHDSI
- OHDSI OHDSI Study Nurture Committee
- OHDSI GitHub R Shiny OHDSI

Chapter A

Chapter B

— — — — —

B.1 ACE

A horizontal row of ten empty square boxes, intended for children to write numbers or letters into. They are arranged in a single line.

- ACE B.1

30

Table B.1: ACE

ID				
1308216	□□□□□□	□□□	□□□	□□□□□□
1310756	□□□□□□□□	□□□	□□	□□□
1331235	□□□□□□	□□□	□□	□□□
1334456	□□□□□□	□□□	□□	□□□
1335471	□□□□□□	□□□	□□	□□□
1340128	□□□□□□	□□□	□□	□□□
1341927	□□□□□□	□□□	□□	□□□
1342439	□□□□□□□□	□□□	□□	□□□
1363749	□□□□□□□□	□□□	□□	□□□
1373225	□□□□□□□□	□□□	□□	□□□

B.2 ACE

□ □ □ □ □ □ □ □ □

- ACE B.2

100

□□□□#1□□□□1□□□□□□□□□□□□□□□□□

A horizontal row of fifteen empty rectangular boxes, intended for children to write their names in, likely as part of a classroom activity or name recognition exercise.

□ □ □ □ #2 □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □ □

A horizontal row of fifteen empty rectangular boxes, likely used for grading student responses.

- 1 B 4 0

A horizontal row of fifteen empty square boxes, likely used for grading student responses.

- 0 7 B 4 1

□ □ □ □ □ □

ACE□□□□□□□□□□□□ (□ B.2)

- **30**
 - **0**

Table B.2: ACE□□□

ID				
1308216	□□□□□□	□□□	□□□	□□□□□□
1310756	□□□□□□□□	□□□	□□	□□□
1331235	□□□□□	□□□	□□	□□□
1334456	□□□□□	□□□	□□	□□□
1335471	□□□□□□	□□□	□□	□□□
1340128	□□□□□□	□□□	□□	□□□
1341927	□□□□□□	□□□	□□	□□□
1342439	□□□□□□□□	□□□	□□	□□□
1363749	□□□□□□□□	□□□	□□	□□□
1373225	□□□□□□□□	□□□	□□	□□□

Table B.3: □□□□□□

□□□□□ID □□□□□□ □□ □□□ □□□□□□
316866 □□□□□□ □□□ □□ □□□

Table B.4: □ □ □ □

ID				
904542	██████████	████	██	██████
907013	██████	████	██	██████
932745	██████	████	██	██████
942350	██████	████	██	██████
956874	██████	████	██	██████
970250	██████████████	████	██	██████

□□□□□ID	□□□□□□	□□	□□□	□□□□□□
974166	□□□□□□□□□□	□□□	□□	□□□
978555	□□□□□□	□□□	□□	□□□
991382	□□□□□	□□□	□□	□□□
1305447	□□□□□	□□□	□□	□□□
1307046	□□□□□□□	□□□	□□	□□□
1307863	□□□□□	□□□	□□	□□□
1308216	□□□□□□	□□□	□□	□□□
1308842	□□□□□□	□□□	□□	□□□
1309068	□□□□□□	□□□	□□	□□□
1309799	□□□□□□	□□□	□□	□□□
1310756	□□□□□□□	□□□	□□	□□□
1313200	□□□□□	□□□	□□	□□□
1314002	□□□□□□	□□□	□□	□□□
1314577	□□□□□□	□□□	□□	□□□
1317640	□□□□□□□	□□□	□□	□□□
1317967	□□□□□□	□□□	□□	□□□
1318137	□□□□□□	□□□	□□	□□□
1318853	□□□□□□	□□□	□□	□□□
1319880	□□□□□□	□□□	□□	□□□
1319998	□□□□□□□	□□□	□□	□□□
1322081	□□□□□□□	□□□	□□	□□□
1326012	□□□□□□	□□□	□□	□□□
1327978	□□□□□□□	□□□	□□	□□□
1328165	□□□□□□	□□□	□□	□□□
1331235	□□□□□	□□□	□□	□□□
1332418	□□□□□□	□□□	□□	□□□
1334456	□□□□□	□□□	□□	□□□
1335471	□□□□□□	□□□	□□	□□□
1338005	□□□□□□□	□□□	□□	□□□
1340128	□□□□□□	□□□	□□	□□□
1341238	□□□□□	□□□	□□	□□□
1341927	□□□□□□	□□□	□□	□□□
1342439	□□□□□□□□□	□□□	□□	□□□
1344965	□□□□□□□	□□□	□□	□□□
1345858	□□□□□□	□□□	□□	□□□
1346686	□□□□□□□	□□□	□□	□□□
1346823	□□□□□□□	□□□	□□	□□□
1347384	□□□□□□□	□□□	□□	□□□
1350489	□□□□□	□□□	□□	□□□
1351557	□□□□□□□	□□□	□□	□□□
1353766	□□□□□□□□	□□□	□□	□□□
1353776	□□□□□□	□□□	□□	□□□
1363053	□□□□□□	□□□	□□	□□□

ID				
1363749	□□□□□□□	□□□	□□□	□□□□□□□
1367500	□□□□□□	□□□	□□□	□□□□
1373225	□□□□□□□□	□□□	□□	□□□□
1373928	□□□□□□□	□□□	□□	□□□□
1386957	□□□□□□□	□□□	□□	□□□□
1395058	□□□□□□□□	□□□	□□	□□□□
1398937	□□□□□□	□□□	□□	□□□□
40226742	□□□□□□□□	□□□	□□	□□□□
40235485	□□□□□□□□	□□□	□□	□□□□

B.3 AMI

- □□□□□□□ B.5□□□□□□□□□□

□ □

Table B.5: □□□□□□□□□□

Table B.6: □□□□□□□□□□

ID				
262	███████████████████	███	███	██████████
9201	██████████	███	██	███
9203	██████████	███	██	███

B.4

A horizontal row of ten empty square boxes, each with a thin black border, intended for children to draw or write in.

A horizontal row of 20 empty square boxes, likely for drawing or writing practice.

□ □ □ □ □ □ □ □ □

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Table B.7: □□□□□

ID 432791

Table B.8: □□□□□□□□□□□

□□□□□ID □□□□□□□ □□ □□□ □□□□□□
262 □□□□□□□□□□□□ □□□ □□ □□□

ID				
9201				
9203				

B.5

□□□□1□□□□1□□□□□□□□□□□□□□□□□□□

- 365 (B.10)

□□□□#2□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□□

□□□#3 ACE □□□

□□□□□□□□□□□□□□□□□□□□□□□□ B.9 □□□□□□□□□□

- **30**
 - **0**

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Table B.9: □□□□□□□□□□□□□□□□

□□□□ID	□□□□□□	□□	□□□	□□□□□□
907013	□□□□□	□□□	□□	□□□
974166	□□□□□□□□□□	□□□	□□	□□□
978555	□□□□□□	□□□	□□	□□□
1395058	□□□□□□	□□□	□□	□□□

Table B.10: □□□□□□

□□□□ID	□□□□□□	□□	□□□	□□□□□□
316866	□□□□□□	□□□	□□	□□□

Table B.11: □□□□□□

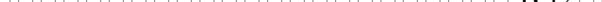
□□□□ID	□□□□□□	□□	□□□	□□□□□□
904542	□□□□□□□	□□□	□□	□□□
907013	□□□□□	□□□	□□	□□□
932745	□□□□□	□□□	□□	□□□
942350	□□□□□	□□□	□□	□□□
956874	□□□□□	□□□	□□	□□□
970250	□□□□□□□□	□□□	□□	□□□
974166	□□□□□□□□□□	□□□	□□	□□□
978555	□□□□□□	□□□	□□	□□□
991382	□□□□□□	□□□	□□	□□□
1305447	□□□□□	□□□	□□	□□□
1307046	□□□□□□□	□□□	□□	□□□
1307863	□□□□□	□□□	□□	□□□
1308216	□□□□□□	□□□	□□	□□□
1308842	□□□□□□	□□□	□□	□□□
1309068	□□□□□□	□□□	□□	□□□
1309799	□□□□□□	□□□	□□	□□□
1310756	□□□□□□□	□□□	□□	□□□
1313200	□□□□□	□□□	□□	□□□
1314002	□□□□□□	□□□	□□	□□□
1314577	□□□□□□	□□□	□□	□□□
1317640	□□□□□□□	□□□	□□	□□□
1317967	□□□□□□	□□□	□□	□□□
1318137	□□□□□□□	□□□	□□	□□□
1318853	□□□□□□	□□□	□□	□□□

ID				
1319880	□□□□□□□	□□□□	□□	□□□
1319998	□□□□□□□□	□□□□	□□	□□□
1322081	□□□□□□□	□□□□	□□	□□□
1326012	□□□□□□□	□□□□	□□	□□□
1327978	□□□□□□□□	□□□□	□□	□□□
1328165	□□□□□□□	□□□□	□□	□□□
1331235	□□□□□□	□□□□	□□	□□□
1332418	□□□□□□□	□□□□	□□	□□□
1334456	□□□□□□	□□□□	□□	□□□
1335471	□□□□□□□	□□□□	□□	□□□
1338005	□□□□□□□□	□□□□	□□	□□□
1340128	□□□□□□□	□□□□	□□	□□□
1341238	□□□□□□	□□□□	□□	□□□
1341927	□□□□□□□	□□□□	□□	□□□
1342439	□□□□□□□□□	□□□□	□□	□□□
1344965	□□□□□□□□	□□□□	□□	□□□
1345858	□□□□□□□	□□□□	□□	□□□
1346686	□□□□□□□□	□□□□	□□	□□□
1346823	□□□□□□□□	□□□□	□□	□□□
1347384	□□□□□□□□	□□□□	□□	□□□
1350489	□□□□□□	□□□□	□□	□□□
1351557	□□□□□□□□	□□□□	□□	□□□
1353766	□□□□□□□□□	□□□□	□□	□□□
1353776	□□□□□□□	□□□□	□□	□□□
1363053	□□□□□□□	□□□□	□□	□□□
1363749	□□□□□□□□	□□□□	□□	□□□
1367500	□□□□□□	□□□□	□□	□□□
1373225	□□□□□□□□	□□□□	□□	□□□
1373928	□□□□□□□	□□□□	□□	□□□
1386957	□□□□□□□	□□□□	□□	□□□
1395058	□□□□□□□□	□□□□	□□	□□□
1398937	□□□□□□	□□□□	□□	□□□
40226742	□□□□□□□□	□□□□	□□	□□□
40235485	□□□□□□□□	□□□□	□□	□□□

B.6

□ □ □ □ □ □ □ □ □

A horizontal row of twelve empty rectangular boxes, intended for children to write their names in, likely as part of a classroom activity.

-  B.12 

□ □ □ □ □

- 136501 B.13
 - 36500 B.14

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Table B.12: □□□□□□□□

ID				
907013	□□□□□	NO	YES	NO
974166	□□□□□□□□□□	NO	YES	NO
978555	□□□□□□	NO	YES	NO
1307863	□□□□□	NO	YES	NO
1308216	□□□□□□	NO	YES	NO
1308842	□□□□□□	NO	YES	NO
1310756	□□□□□□□	NO	YES	NO
1317640	□□□□□□□	NO	YES	NO
1318137	□□□□□□	NO	YES	NO
1318853	□□□□□□	NO	YES	NO
1319880	□□□□□□	NO	YES	NO
1326012	□□□□□□	NO	YES	NO
1328165	□□□□□□	NO	YES	NO
1331235	□□□□□	NO	YES	NO
1332418	□□□□□□	NO	YES	NO
1334456	□□□□□	NO	YES	NO
1335471	□□□□□□	NO	YES	NO
1340128	□□□□□□	NO	YES	NO
1341927	□□□□□□	NO	YES	NO
1342439	□□□□□□□□	NO	YES	NO
1346686	□□□□□□□□	NO	YES	NO

□□□□□ID	□□□□□□□	□□	□□□	□□□□□□□
1347384	□□□□□□□	NO	YES	NO
1351557	□□□□□□□	NO	YES	NO
1353776	□□□□□□□	NO	YES	NO
1363749	□□□□□□□	NO	YES	NO
1367500	□□□□□□	NO	YES	NO
1373225	□□□□□□□	NO	YES	NO
1395058	□□□□□□□	NO	YES	NO
40226742	□□□□□□□	NO	YES	NO
40235485	□□□□□□□	NO	YES	NO

Table B.13: □□□□□□□

□□□□□ID	□□□□□□□	□□	□□□	□□□□□□□
904542	□□□□□□□	NO	YES	NO
907013	□□□□□	NO	YES	NO
932745	□□□□□	NO	YES	NO
942350	□□□□□	NO	YES	NO
956874	□□□□□	NO	YES	NO
970250	□□□□□□□□□	NO	YES	NO
974166	□□□□□□□□□□	NO	YES	NO
978555	□□□□□□	NO	YES	NO
991382	□□□□□	NO	YES	NO
1305447	□□□□□	NO	YES	NO
1307046	□□□□□□□	NO	YES	NO
1307863	□□□□□	NO	YES	NO
1308216	□□□□□□	NO	YES	NO
1308842	□□□□□□	NO	YES	NO
1309068	□□□□□□	NO	YES	NO
1309799	□□□□□□	NO	YES	NO
1310756	□□□□□□□	NO	YES	NO
1313200	□□□□□	NO	YES	NO
1314002	□□□□□□	NO	YES	NO
1314577	□□□□□□	NO	YES	NO
1317640	□□□□□□□	NO	YES	NO
1317967	□□□□□□	NO	YES	NO
1318137	□□□□□□	NO	YES	NO
1318853	□□□□□□	NO	YES	NO
1319880	□□□□□□	NO	YES	NO
1319998	□□□□□□□	NO	YES	NO
1322081	□□□□□□□	NO	YES	NO
1326012	□□□□□□□	NO	YES	NO
1327978	□□□□□□□	NO	YES	NO

ID				
1328165	□□□□□□□	NO	YES	NO
1331235	□□□□□□	NO	YES	NO
1332418	□□□□□□□	NO	YES	NO
1334456	□□□□□□	NO	YES	NO
1335471	□□□□□□□	NO	YES	NO
1338005	□□□□□□□□	NO	YES	NO
1340128	□□□□□□□	NO	YES	NO
1341238	□□□□□□	NO	YES	NO
1341927	□□□□□□□	NO	YES	NO
1342439	□□□□□□□□□	NO	YES	NO
1344965	□□□□□□□□	NO	YES	NO
1345858	□□□□□□□	NO	YES	NO
1346686	□□□□□□□□	NO	YES	NO
1346823	□□□□□□□□	NO	YES	NO
1347384	□□□□□□□□	NO	YES	NO
1350489	□□□□□□	NO	YES	NO
1351557	□□□□□□□□	NO	YES	NO
1353766	□□□□□□□□□	NO	YES	NO
1353776	□□□□□□□	NO	YES	NO
1363053	□□□□□□□	NO	YES	NO
1363749	□□□□□□□	NO	YES	NO
1367500	□□□□□□	NO	YES	NO
1373225	□□□□□□□□	NO	YES	NO
1373928	□□□□□□□	NO	YES	NO
1386957	□□□□□□□	NO	YES	NO
1395058	□□□□□□□□	NO	YES	NO
1398937	□□□□□□	NO	YES	NO
40226742	□□□□□□□□	NO	YES	NO
40235485	□□□□□□□□	NO	YES	NO

Table B.14: □□□□□□

□□□□□ID □□□□□□ □□ □□□ □□□□□□
316866 □□□□□□ NO YES NO

B.7 3

B.8 ACE

- ACE□□□□□ B.15□□□□□□□□

ACE□□□□□□□□□□ B.15□

- **30**
 - **0**

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Table B.15: ACE□□□

ID				
1308216	□□□□□□	NO	YES	NO
1310756	□□□□□□□□	NO	YES	NO
1331235	□□□□□□	NO	YES	NO
1334456	□□□□□□	NO	YES	NO
1335471	□□□□□□□	NO	YES	NO
1340128	□□□□□□	NO	YES	NO
1341927	□□□□□□□	NO	YES	NO
1342439	□□□□□□□□□	NO	YES	NO
1363749	□□□□□□□	NO	YES	NO
1373225	□□□□□□□□	NO	YES	NO

B.9 ARB

□□□□□□□□□

ID				
1308842	██████████	NO	YES	NO
1317640	███████████	NO	YES	NO
1346686	███████████	NO	YES	NO
1347384	███████████	NO	YES	NO
1351557	███████████	NO	YES	NO
1367500	███████	NO	YES	NO
40226742	███████████	NO	YES	NO
40235485	███████████	NO	YES	NO

B.10

B.8 B.17 ACE B.15

Table B.17: □□□□□□□□□□□□□□□□□□□□□□□□

ID				
907013	□□□□□	NO	YES	NO
974166	□□□□□□□□□□	NO	YES	NO
978555	□□□□□□	NO	YES	NO
1395058	□□□□□□□	NO	YES	NO

B.11

DCCB

ID				
1318137		NO	YES	NO
1318853		NO	YES	NO
1319880		NO	YES	NO
1326012		NO	YES	NO

ID				
1332418		NO	YES	NO
1353776		NO	YES	NO

B.12 **NDCCB**

B.8 NDCCB B.19 ACE B.15

ID					
1307863		NO	YES	NO	
1328165		NO	YES	NO	

B.13

□□□□□ B.8 □□□□□□□ β □□□ (□ B.20) □ ACE □□□ (□ B.15) □□□□□□□□□□□

Table B.20: β □□□

ID				
1307046	□□□□□□□	NO	YES	NO
1313200	□□□□□□	NO	YES	NO
1314002	□□□□□□□	NO	YES	NO
1314577	□□□□□□□	NO	YES	NO
1319998	□□□□□□□	NO	YES	NO
1322081	□□□□□□□	NO	YES	NO
1327978	□□□□□□□	NO	YES	NO
1338005	□□□□□□□	NO	YES	NO
1345858	□□□□□□□	NO	YES	NO
1346823	□□□□□□□	NO	YES	NO
1353766	□□□□□□□□□	NO	YES	NO
1386957	□□□□□□□	NO	YES	NO

B.14

ACE □□□□□ B.8 □□□□□□□□□□□□□□□□□ (□ B.21) □*ACE*□□□ (□ B.15)
□□□□□□□□□□□□□□□

Table B.21: □□□□□□

ID				
932745	□□□□□□	NO	YES	NO
942350	□□□□□□	NO	YES	NO
956874	□□□□□□	NO	YES	NO

B.15

Table B.22: □□□□□□□□□□

ID				
904542		NO	YES	NO
991382		NO	YES	NO

B.16 1

ACE□□□□□ B.8 □□□□□□□□ *αI*□□□(□ B.23) □ *ACE*□□□(□ B.15)□□□□□□□□□□

Table B.23: α_1 □□□

ID				
1341238	□□□□□	NO	YES	NO
1350489	□□□□□	NO	YES	NO
1363053	□□□□□□	NO	YES	NO

Chapter C

— —

C.1 ACE

ID	
434165	□□□□□□□□□□
436409	□□□□
199192	□□□□□□□□□□□□□□□□□□□□/□□□□□□□□
4088290	□□□□
4092879	□□□□
44783954	□□□□
75911	□□□□□□□
137951	□□□□□□□□□
77965	□□□□□□□
376707	□□□□□
4103640	□□□
73241	□□□□□□□□□□□□□
133655	□□□□□
73560	□□□□□
434327	□□□□
4213540	□□□□□□□□□□
140842	□□□□□□□□□
81378	□□□□□□□□□
432303	□□□□□□□
4201390	□□□□□□□
46269889	□□□□□□□□□□□□□

□□□□□ID	□□□□□□□□□
134438	□□□□□□
78619	□□□□□
201606	□□□□□
76786	□□□□□□□
4115402	□□□□
45757370	□□□□□□□□□
433111	□□□□□
433527	□□□□□
4170770	□□□□□
4092896	□□□□□□□
259995	□□□□□□□
40481632	□□□□□□□□□
4166231	□□□□□
433577	□□□
4231770	□□□□□□□
440329	□□□□□□□□□□□□□
4012570	□□□□□□□□□
4012934	□□□□□□□□□
441788	□□□□□□□□□□□□□
4201717	□□□□□□□
374375	□□□□
4344500	□□□□□□□□□□□□□□
139099	□□□
444132	□□□
196168	□□□□
432593	□□□□□□□
434203	□□□□□□□
438329	□□□□□□□□□
195873	□□□□
4083487	□□□□□□□
4103703	□□□□□□□□□
4209423	□□□□□□□
377572	□□□□□□□□□□□
40480893	□□□□□□□□□□□□□□□
136368	□□□□□□□□□□□
140648	□□□□□□□□□□□
438130	□□□□□□□
4091513	□□
4202045	□□□□□□□□□□□□
373478	□□
46286594	□□□□□□□□□□□□□
439790	□□□□
81634	□□□□

□□□□ID	□□□□□□□
380706	□□□
141932	□□□□□□
36713918	□□□□□□□□□
443172	□□□□□□□□□□□□
81151	□□□□□
72748	□□□□□□□□□□□
378427	□□□□
437264	□□□□□□□□
194083	□□□□□□□□□
140641	□□□□□
440193	□□□□
4115367	□□□□

Chapter D

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16. □□□□

Chapter E

— —

E.1

4.1

E.1

Table E.1: PERSON

□□□□	□	□□
PERSON_ID	2	□□□□□□
GENDER_CONCEPT_ID	8507	□□□□□□□ ID □ 8507□
YEAR_OF_BIRTH	1974	
MONTH_OF_BIRTH	8	
DAY_OF_BIRTH	4	
BIRTH_DATETIME	1974-08-04 00:00:00	□□□□□□□□□□00□00:00:00□□□□
DEATH_DATETIME	NULL	
RACE_CONCEPT_ID	8516	□□□□□□□□□□□□□□□□ ID □ 8516□
ETHNICITY_CONCEPT_ID	38003564	38003564
LOCATION_ID		□□□□□□□□□□□□□□
PROVIDER_ID		□□□□□□□
CARE_SITE		□□□□□□□□□□□□
PERSON_SOURCE_VALUE	NULL	□□□□□□□□□□
GENDER_SOURCE_VALUE	Man	□□□□□□□□□□□□

□	□	□
GENDER_SOURCE_	0	
CONCEPT_ID		
RACE_SOURCE_	African American	□□□□□□□□□□□□□
VALUE		
RACE_SOURCE_	0	
CONCEPT_ID		
ETHNICITY_SOURCE_	NULL	
VALUE		
ETHNICITY_SOURCE_	0	
CONCEPT_ID		

4.2

E.2

Table E.2: OBSERVATION_PERIOD

□□□□	□	□□
OBSERVATION_	2	□□□□□□□
PERIOD_ID		
PERSON_ID	2	□□□PERSON□□□□□□□
OBSERVATION_PERIOD_	2015-01-01	□□□□□□□
START_DATE		
OBSERVATION_PERIOD_	2019-07-01	□□□□□□□□□□□□□□□
END_DATE		
PERIOD_TYPE_	44814722	44814724
CONCEPT_ID		□□□□□□□□□□□□□□□

4.3

E.3

Table E.3: DRUG EXPOSURE

□□□□	□	□□
DRUG_EXPOSURE_ID	1001	□□□□□□
PERSON_ID	2	□□□PERSON□□□□□□□□□□□□□□□□□□
DRUG_CONCEPT_ID	19078461	□□□□□NDC□□□□□□□□□□□□
		19078461
		□□□□□□□□□□
DRUG_EXPOSURE_	2019-05-01	□□□□□□□□□□
START_DATE		

□□□□	□	□□
DRUG_EXPOSURE_	2019-05-01 00:00:00	□□□□□□□□□0□□□□□
START_DATETIME		
DRUG_EXPOSURE_	2019-05-31	□□□ + □□□□□□□□□□
END_DATE		
DRUG_EXPOSURE_	2019-05-31 00:00:00	□□□□□□□□□0□□□□□
END_DATETIME		
VERBATIM_END_DATE	NULL	□□□□□□□□□□
DRUG_TYPE_	38000177	38000177
CONCEPT_ID		□□□□□□□□□□□□□□
STOP_REASON	NULL	
REFILLS	NULL	
QUANTITY	NULL	□□□□□□□□□□
DAYS_SUPPLY	30	□□□□□□□□□□□□□□
SIG	NULL	□□□□□□□□□□
ROUTE_CONCEPT_ID	4132161	4132161 □□□□□□□□□□
LOT_NUMBER	NULL	□□□□□□□□□□
PROVIDER_ID	NULL	□□□□□□□□□□
VISIT_OCCURRENCE_	NULL	□□□□□□□□□□□□□□□□□□□□□□
ID		
VISIT_DETAIL_ID	NULL	
DRUG_SOURCE_	76168009520	□□□□□NDC□□□□
VALUE		
DRUG_SOURCE_	583945	583945
CONCEPT_ID		□□□□□□□□□□□□□□
(ND-		
C□□□□76168009520□)		□
ROUTE_SOURCE_	NULL	
VALUE		

4.4

□□□□□□□□□□□□ CONDITION OCCURRENCE □□□□□□□□□□□□

```
library(DatabaseConnector)
connection <- connect(connectionDetails)
sql <- "SELECT *
FROM @cdm.condition_occurrence
WHERE condition_concept_id = 192671;"

result <- renderTranslateQuerySql(connection, sql, cdm = "main")
head(result)
```

CONDITION_OCCURRENCE_ID PERSON_ID CONDITION_CONCEPT_ID

```

## 1           4657      273        192671 ...
## 2           1021       61        192671 ...
## 3          5978      351        192671 ...
## 4          9798      579        192671 ...
## 5          9301      549        192671 ...
## 6          1997      116        192671 ...

```

□ □ 4.5

□□□□□□□□□□ CONDITION OCCURRENCE □□□□ CONDITION SOURCE VALUE

```
sql <- "SELECT *
FROM @cdm.condition_occurrence
WHERE condition_source_value = 'K92.2';"

result <- renderTranslateQuerySql(connection, sql, cdm = "main")
head(result)
```

```
##   CONDITION_OCCURRENCE_ID PERSON_ID CONDITION_CONCEPT_ID ...
## 1                   4657      273           192671 ...
## 2                   1021       61           192671 ...
## 3                   5978      351           192671 ...
## 4                   9798      579           192671 ...
## 5                   9301      549           192671 ...
## 6                  1997     116           192671 ...
```

4.6

OBSERVATION PERIOD

```
library(DatabaseConnector)
connection <- connect(connectionDetails)
sql <- "SELECT *
FROM @cdm.observation_period
WHERE person_id = 61;"

renderTranslateQuerySql(connection, sql, cdm = "main")
```

```
##    OBSERVATION_PERIOD_ID PERSON_ID OBSERVATION_PERIOD_START_DATE ...
## 1                      61        61                     1968-01-21 ...
```

E.2

5.1

□□□□ ID 192671 (“□□□□”)

5.2

ICD-10CM□□□□

- K29.91 “□□□□□□□□□□□□□□□□□□”
 - K92.2 “□□□□□□□□□□□□”

ICD-9CM□□□□

- 578 “□□□□”
 - 578.9 “□□□□□□□□□□□□□□”

5.3

MedDRA□□□□□

- “□□□□” (□□□□ ID 35707864)
 - “□□□” (□□□□ ID 35707858)

E.3

6.1

- A) CDM
 - B) ETL
 - C) ETL
 - D)

6.2

□□□	□	□□
PERSON_ID	A123B456	□□□□□□□□□□□□□□□□□□□□□□□□
GENDER_CONCEPT_ID	8532	□□□□□□□□□□□□□□□□□□□□□□□□
YEAR_OF_BIRTH	NULL	□□□□□□□□□□□□□□□□□□□□□□□□
MONTH_OF_BIRTH	NULL	□□□□□□□□□□□□□□□□□□□□□□□□
DAY_OF_BIRTH	NULL	□□□□□□□□□□□□□□□□□□□□□□□□
RACE_CONCEPT_ID	0	□□□WHITE□□□□□□□□8527□□□
ETHNICITY_CONCEPT_ID	8527	□□□□□□□□□□□□□□□□□□□□□□□□0□
PERSON_SOURCE_VALUE	A123B456	
GENDER_SOURCE_VALUE	F	
RACE_SOURCE_VALUE	WHITE	
ETHNICITY_SOURCE_VALUE	NONE PROVIDED	

6.3

□	□	□
VISIT_OCCURRENCE_ID	1	
PERSON_ID	11	
VISIT_START_DATE	2004-09-26	
VISIT_END_DATE	2004-09-30	
VISIT_CONCEPT_ID	9201	
VISIT_SOURCE_VALUE	inpatient	

E.4

7.1

1. □□□□
 2. □□□□□□□□□□
 3. □□□□□□□□□□

□ □ 7.2

E.5 SQL R

9.1

```
library(DatabaseConnector)
connection <- connect(connectionDetails)
sql <- "SELECT COUNT(*) AS person_count
FROM @cdm.person;"

renderTranslateQuerySql(connection, sql, cdm = "main")
```

```
##      PERSON_COUNT  
## 1      2694
```

9.2

□□□□□□□□□□□□□□□□□□ 1 □□□□□□□□□□□□□□□□□□ DRUG EXPOSURE □□□□□□

```
library(DatabaseConnector)
connection <- connect(connectionDetails)
sql <- "SELECT COUNT(DISTINCT(person_id)) AS person_count
FROM @cdm.drug_exposure
INNER JOIN @cdm.concept_ancestor
    ON drug_concept_id = descendant_concept_id
INNER JOIN @cdm.concept ingredient
    ON ancestor_concept_id = ingredient.concept_id
WHERE LOWER(ingredient.concept_name) = 'celecoxib'
    AND ingredient.concept_class_id = 'Ingredient'
    AND ingredient.standard_concept = 'S';"

renderTranslateQuerySql(connection, sql, cdm = "main")
```

```
##      PERSON_COUNT  
## 1          1844
```

```
library(DatabaseConnector)
connection <- connect(connectionDetails)

sql <- "SELECT COUNT(DISTINCT(person_id)) AS person_count
FROM @cdm.drug_era
INNER JOIN @cdm.concept ingredient
  ON drug_concept_id = ingredient.concept_id
WHERE LOWER(ingredient.concept_name) = 'celecoxib'
  AND ingredient.concept_class_id = 'Ingredient'
  AND ingredient.standard_concept = 'S';"

renderTranslateQuerySql(connection, sql, cdm = "main")
```

```
##      PERSON_COUNT  
## 1      1844
```

9.3

CONDITION OCCURRENCE

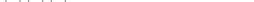
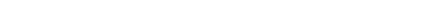
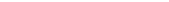
```
library(DatabaseConnector)
connection <- connect(connectionDetails)
sql <- "SELECT COUNT(*) AS diagnose_count
FROM @cdm.drug_era
INNER JOIN @cdm.concept ingredient
    ON drug.concept_id = ingredient.concept_id"
```

```
    INNER JOIN @cdm.condition_occurrence
        ON condition_start_date >= drug_era_start_date
           AND condition_start_date <= drug_era_end_date
    INNER JOIN @cdm.concept_ancestor
        ON condition_concept_id = descendant_concept_id
WHERE LOWER(ingredient.concept_name) = 'celecoxib'
    AND ingredient.concept_class_id = 'Ingredient'
    AND ingredient.standard_concept = 'S'
    AND ancestor_concept_id = 192671;"
```

```
##      DIAGNOSE_COUNT  
## 1          41
```

E.6

□ □ 10.1

- 
 - 16 
 -  365 

E.1

Cohort Entry Events ?

Events having any of the following criteria:

[+ Add Initial Event ▾](#)
[Delete Criteria](#)

a drug era of diclofenac ▾

X for the first time in the person's history

X with age in years at era start Greater or Equal To ▾ 16

with continuous observation of at least 365 ▾ days before and 0 ▾ days after event index date

Limit initial events to: earliest event ▾ per person.

[Restrict initial events](#)

E.2 Diclofenac

Concept Set Expression		Included Concepts (11473)	Included Source Codes		Export	Import		
Name:								
diclofenac								
Show	25	entries						
						Search: <input type="text"/>		
Showing 1 to 1 of 1 entries						Previous <input type="button" value="1"/> Next <input type="button" value="2"/>		
Concept Id	Concept Code	Concept Name	Domain	Standard Concept Caption	Exclude	Descendants	Mapped	
1124300	3355	Diclofenac	Drug	Standard	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	

Figure E.2: □□□□□□□□□□□□□□□□□□□□□□□□

Inclusion Criteria ?

New inclusion criteria
Without prior exposure to any NSAID
Copy Delete

1. Without prior exposure to any NSAID

Excluding subjects with prior exposure to any NSAID

Excluding subjects with prior exposure to any NSAID

having all of the following criteria:

+ Add criteria to group...

with exactly 0 using all occurrences of:

a drug exposure of NSAIDs ▼

where event starts between All days Before 1 days Before

index start date add additional constraint

restrict to the same visit occurrence

allow events from outside observation period

Delete Criteria

Limit qualifying events to: earliest event per person.

Figure E.3: NSAID□□□□□□□□□□□□

Concept Set Expression		Included Concepts 23112	Included Source Codes		Export	Import			
Name: <input type="text" value="NSAIDs"/>									
Show 25 entries Search: <input type="text"/>									
Showing 1 to 1 of 1 entries Previous 1 Next									
	Concept Id	Concept Code	Concept Name	Domain	Standard Concept Caption	Exclude	Descendants	Mapped	
	21603933	M01A	ANTIINFLAMMATORY AND ANTIRHEUMATIC PRODUCTS, NON-STEROIDS	Drug	Classification	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	

Figure E.4: NSAIDs

E.5

Inclusion Criteria

New inclusion criteria

Without prior diagnosis of cancer Copy **Delete**

1. Without prior exposure to any NSAID
Excluding subjects with prior exposure to any NSAID

Excluding subjects with prior cancer diagnosis

having **all** of the following criteria:

+ Add criteria to group... **Delete Criteria**

2. Without prior diagnosis of cancer
Excluding subjects with prior cancer diagnosis

with **exactly** 0 occurrences of:
 a condition occurrence of **Broad malignancies** + Add attribute...

where **event starts** between **All** days **Before** and **0** days **Before**
index start date [add additional constraint](#)

restrict to the same visit occurrence
 allow events from outside observation period

Limit qualifying events to: **earliest event** per person.

Figure E.5: □□□□□□□□□□□□□□□□□□

E.6

Concept Set Expression	Included Concepts 4401	Included Source Codes	Export	Import			
Name:							
Broad malignancies							
Show	25 ▾ entries	Search: <input type="text"/>					
Showing 1 to 1 of 1 entries							
Concept Id	Concept Code	Concept Name	Domain	Standard Concept Caption	Exclude	Descendants	Mapped
443392	363346000	Malignant neoplastic disease	Condition	Standard	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>

Figure E.6: □□□□□□□□□□□□□□□□□□

30 E 7

□ □ 10.2

SQL 2

```
library(DatabaseConnector)
connection <- connect(connectionDetails)
sql <- "SELECT person_id AS subject_id,
    condition_start_date AS cohort_start_date
INTO #diagnoses
FROM @cdm.condition_occurrence
WHERE condition_concept_id IN (
    SELECT descendant_concept_id
    FROM @cdm.concept_ancestor
    WHERE ancestor_concept_id = 4648
    )"
```

Cohort Exit

Event Persistence:
Event will persist until: ▾

Continuous Exposure Persistence:
Specify a concept set that contains one or more drugs. A drug era will be derived from all drug exposure events for any of the drugs within the concept set, using the specified persistence window as a maximum allowable gap in days between successive exposure events and adding a specified surveillance window to the final exposure event. If no exposure event end date is provided, then an exposure event end date is inferred to be event start date + days supply in cases when days supply is available or event start date + 1 day otherwise. This event persistence assures that the cohort end date will be no greater than the drug era end date.

Concept set containing the drug(s) of interest: ▾

- Persistence window: allow for a maximum of days between exposure records when inferring the era of persistence exposure
- Surveillance window: add days to the end of the era of persistence exposure as an additional period of surveillance prior to cohort exit.

Censoring Events:

Exit Cohort based on the following criteria:

No censoring events selected.

Figure E.7: □□□□□□□□□□

```
        WHERE ancestor_concept_id = 4329847 --  
    )  
    AND condition_concept_id NOT IN (  
        SELECT descendant_concept_id  
        FROM @cdm.concept_ancestor  
        WHERE ancestor_concept_id = 314666 --  
    );"  
  
renderTranslateExecuteSql(connection, sql, cdm = "main")
```

ER COHORT DEFINITION ID

```
sql <- "INSERT INTO @cdm.cohort (
  subject_id,
  cohort_start_date,
  cohort_definition_id
)
SELECT subject_id,
  cohort_start_date,
  CAST (1 AS INT) AS cohort_definition_id
FROM #diagnoses
INNER JOIN @cdm.visit_occurrence
  ON subject_id = person_id
    AND cohort_start_date >= visit_start_date
    AND cohort_start_date <= visit_end_date
WHERE visit_concept_id IN (9201, 9203, 262); --      ;"
```

renderTranslateExecuteSql(connection, sql, cdm = "main")

VISIT_OCCURRENCE_ID: 1

```
sql <- "TRUNCATE TABLE #diagnoses;  
DROP TABLE #diagnoses;"  
  
renderTranslateExecuteSql(connection, sql)
```

E.7

11.1

ATLAS Data Sources
E.8

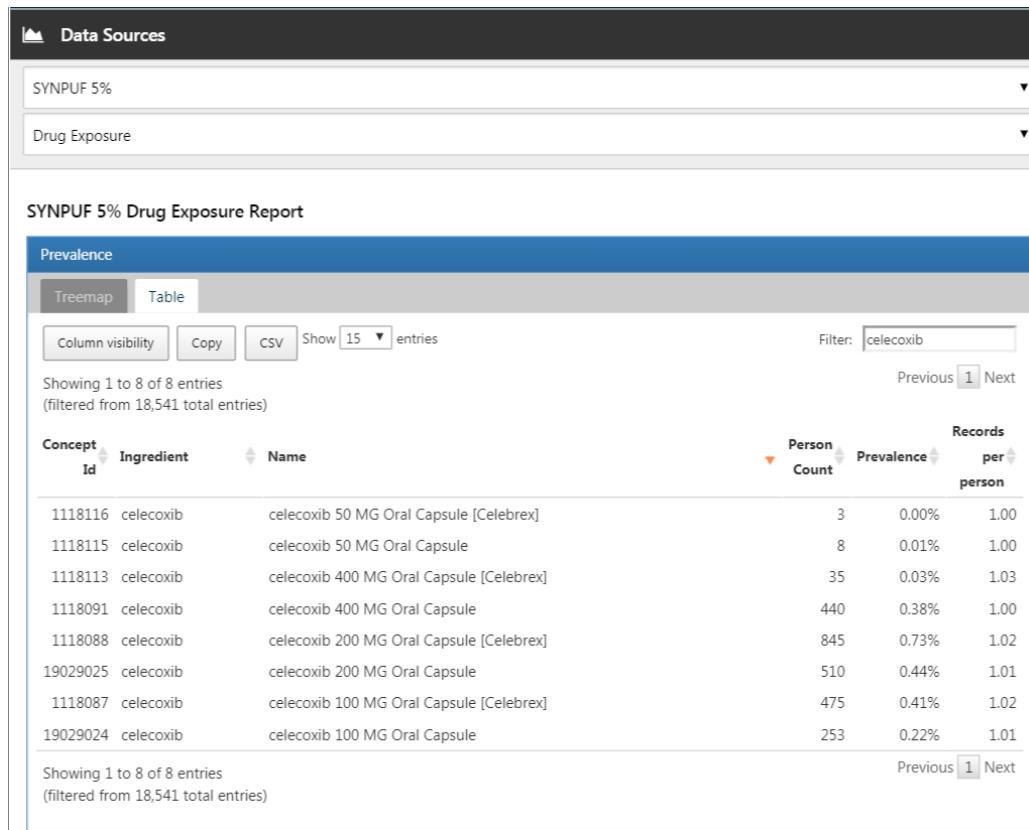


Figure E.8: □□□□□□□□□□□□

□ □ 11.2

Cohort Definitions

Search

E.9

← Celecoxib new users ➔ Celecoxib

Q Search

Search Import

celexcoxib

[Advanced Options](#)

Showing 1 to 1 of 1 entries

	Id	Code	Name	Class	RC	DRC	Domain	Vocabulary
	1118084	140587	celexcoxib	Ingredient	2,587	5,184	Drug	RxNorm

Showing 1 to 1 of 1 entries

Vocabulary

- RxNorm Extension (1376)
- NDC (1337)
- SPL (449)
- DPD (167)
- [SNOMED \(75\)](#)

Class

- Ingredient (7)**
- Clinical Drug Form (5)
- Clinical Drug Comp (5)
- Lab Test (5)
- [Other \(5\)](#)

Domain

- Drug (3570)
- Measurement (18)
- Observation (1)
- Meas Value (1)

Standard Concept

- Non-Standard (1831)
- Standard (1292)**
- Classification (467)

Figure E.9: □□□□□□□□□□□□□□□□□□□□□□□□□□□□□□

E.9 +



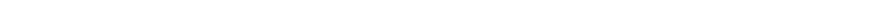
 Characterizations

E.11 E.12

□□□□□□□□□□□□□□□□ 1 □□□□□□□□□□□□□□□□□□□□□□

E.12 □ □ □ □

11.3

 Cohort Definitions  E.13 

New Cohort Definition

Celecoxib new users Save  Cancel 

- [Definition !\[\]\(fb19c0689fa6de7a5107037a8c7e9217_img.jpg\)](#)
- [Concept Sets](#)
- [Generation](#)
- [Reporting](#)
- [Export](#)

enter a cohort definition description here

Cohort Entry Events

Events having any of the following criteria:

a drug era of Celecoxib   Add Initial Event 

 Delete Criteria

 for the first time in the person's history

with continuous observation of at least 365  days before and 0  days after event index date
 Limit initial events to: earliest event  per person.

Restrict initial events

Figure E.10: □□□□□□□□□□□□□□□□□□□□□□□□□□□□□

E.15 + 

E.16  Incidence Rates

E.17 1

E.8

□ □ 12.1

```
library(CohortMethod)
nsaids <- c(1118084, 1124300) #
covSettings <- createDefaultCovariateSettings(
  excludedCovariateConceptIds = nsaids,
  addDescendantsToExclude = TRUE)

#
cmData <- getDbCohortMethodData(
  connectionDetails = connectionDetails,
```

New Characterization

Celecoxib new users characterization

Design Executions Utilities

Cohort characterization is defined as the process of generating cohort level descriptive summary statistics from person level covariate data. Summary statistics of these person level covariates may be count, mean, sd, var, min, max, median, range, and quantiles. In addition, covariates during a period may be stratified into temporal units of time for time-series analysis such as fixed intervals of time relative to cohort_start_date (e.g. every 7 days, every 30 days etc.), or in absolute calendar intervals such as calendar-week, calendar-month, calendar-quarter, calendar-year.

Cohort definitions

Import

Show 10 entries Search:

ID	Name	Actions
1771701	Celecoxib new users	Edit cohort Remove

Showing 1 to 1 of 1 entries Previous 1 Next

Feature analyses

Import

Show 10 entries Search:

ID	Name	Description	Actions
15	Drug Group Era Any Time Prior	One covariate per drug rolled up to ATC groups in the drug_era table overlapping with any time prior to index.	Remove
27	Condition Group Era Any Time Prior	One covariate per condition era rolled up to groups in the condition_era table overlapping with any time prior to index.	Remove

Showing 1 to 2 of 2 entries Previous 1 Next

Figure E.11: □□□□□□

Characterization #69

Celecoxib new users characterization

Design Executions Utilities

Executions > Reports for SYNPUF 5%

Date: 08/23/2019 12:53 PM Design: -1840810470 Results: 2 reports

Filter panel

Cohorts	Analyses	Domains
Celecoxib new users	Condition Group Era Any Time P	Condition, Drug

CONDITION / Condition Group Era Any Time Prior

Export Show 10 entries Search:

Covariate	Explore	Concept ID	Count	Pct
Pain	Explore	4329041	1,140	78.62%
Pain finding at anatomical site	Explore	4132926	1,135	78.28%
Inflammation of specific body systems	Explore	4178818	1,135	78.28%
Arthropathy	Explore	73553	1,122	77.38%

Figure E.12: □□□□□□□

◀ GI bleed ▶ GI bleed

Search

Search Import

Gastrointestinal hemorrhage

Advanced Options

Column visibility Copy CSV Show 15 entries Filter: Previous 1 2 Next

Showing 1 to 15 of 25 entries

Vocabulary	Id	Code	Name	Class	RC	DRC	Domain	Vocabulary
SNOMED (17)	192671	74474003	Gastrointestinal hemorrhage	Clinical Finding	919	37,144	Condition	SNOMED
ICD10CM (2)	4338544	87763006	Lower gastrointestinal hemorrhage	Clinical Finding	0	15,617	Condition	SNOMED
ICD9CM (2)								
DRG (2)								
NIDERT (1)								
Class	4100660	27719009	Acute gastrointestinal hemorrhage	Clinical Finding	0	9,852	Condition	SNOMED
Clinical Finding (17)								

Figure E.13: □□□□□□□□□□□□□□□□□□□

Concept Set Expression Included Concepts (191) Included Source Codes Export Import

Name: GI bleed

Show 25 entries Search: Previous | 1 | Next

Showing 1 to 1 of 1 entries

Concept Id	Concept Code	Concept Name	Domain	Standard Concept Caption	Exclude	Descendants	Mapped
192671	74474003	Gastrointestinal hemorrhage	Condition	Standard	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>

■ Classification ■ Non-Standard ■ Standard

Figure E.14: □□□□□□□□□□□□□□□□□□□□□□□□

New Cohort Definition

GI bleed

Definition Concept Sets Generation Reporting Export

enter a cohort definition description here

Cohort Entry Events

Events having any of the following criteria:

a condition occurrence of GI bleed ▼ + Add Initial Event

+ Add attribute...▼ Delete Criteria

with continuous observation of at least 0 days before and 0 days after event index date

Limit initial events to: earliest event per person.

Restrict initial events

Figure E.15: □□□□□□□□□□□□□□□□□□

New Incidence Rate Analysis

Incidence of GI bleed after celecoxib initiation

Definition Concept Sets Generation Utilities

Study Cohorts

Target Cohorts: #1771701:Celecoxib new users Add Target Cohort

Outcome Cohorts: #1771702:GI bleed Add Outcome Cohort

Time At Risk

Time at risk defines the time window relative to the cohort start or end date with an offset to consider the person 'at risk' of the outcome.

- Time at risk starts with start date plus 0 days.
- Time at risk ends with start date plus 1095 days.

No study window defined. Add Study Window

Stratify Criteria: You can provide optional stratification criteria to the analysis that will divide the population into unique groups based on their satisfied criteria.

New stratify criteria Please select a qualifying inclusion criteria to edit

Figure E.16: □□□□□□

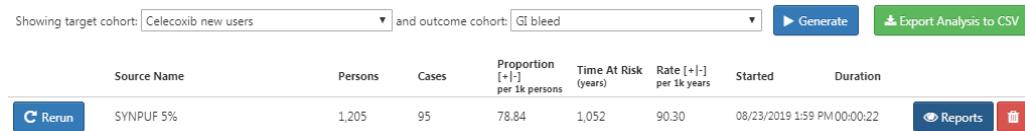


Figure E.17: □□□□□

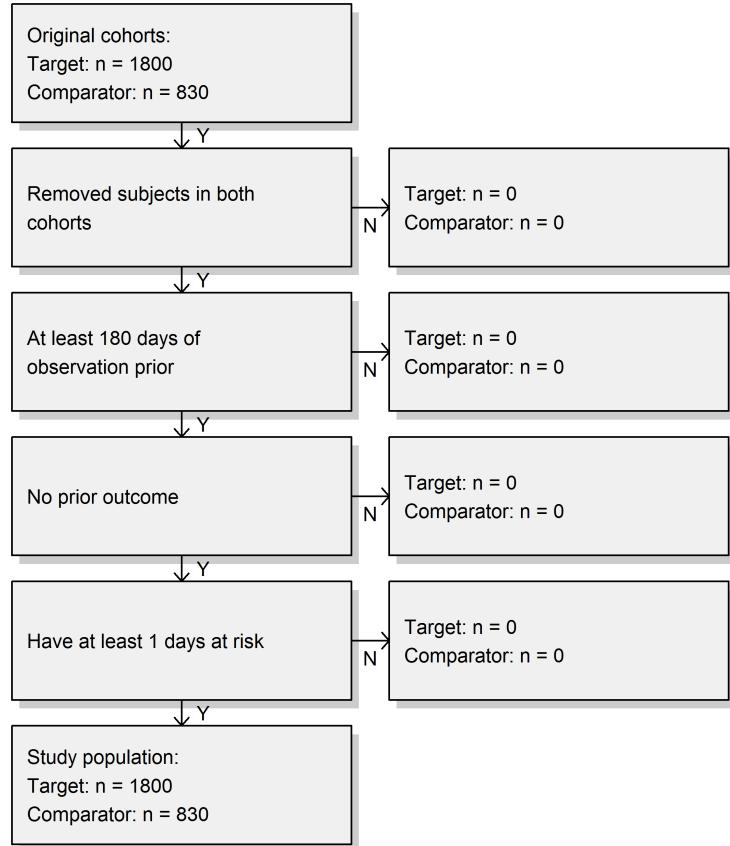
```
cdmDatabaseSchema = "main",
targetId = 1,
comparatorId = 2,
outcomeIds = 3,
exposureDatabaseSchema = "main",
exposureTable = "cohort",
outcomeDatabaseSchema = "main",
outcomeTable = "cohort",
covariateSettings = covSettings)
summary(cmData)
```

```
## CohortMethodData
##
##      ID 1
##      ID 2
##      ID 3
##
##          1800
##          830
##
##          479
##          479
##
##          389
##          26923
```

□ □ 12.2

```
studyPop <- createStudyPopulation(  
    cohortMethodData = cmData,  
    outcomeId = 3,  
    washoutPeriod = 180,  
    removeDuplicateSubjects = "remove all",  
    removeSubjectsWithPriorOutcome = TRUE,
```

```
riskWindowStart = 0,  
startAnchor = "cohort start",  
riskWindowEnd = 99999)  
drawAttritionDiagram(studyPop)
```



□ □ 12.3

Cox

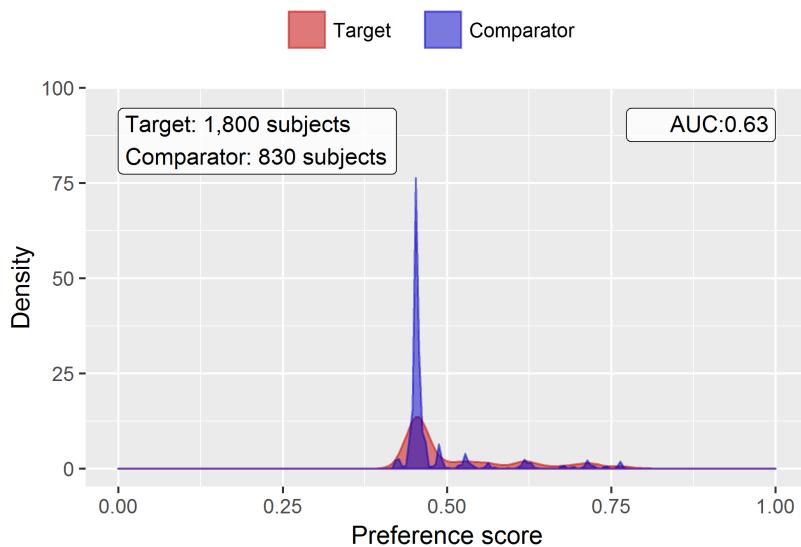
```
model <- fitOutcomeModel(population = studyPop,
                           modelType = "cox")
model
```

```
##      COX  
##      FALSE
```

```
## FALSE
## FALSE
## OK
##
## .95 .95
## 1.34612 1.10065 1.65741 0.29723 0.1044
```

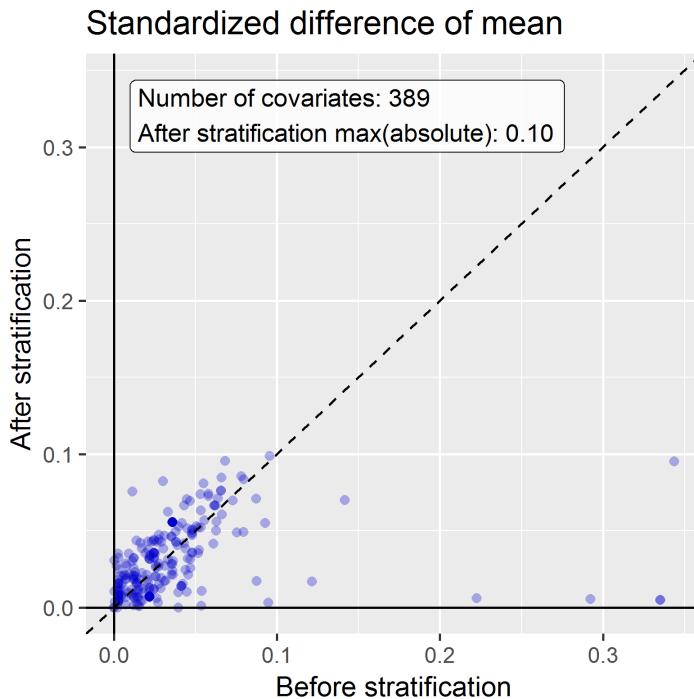
□ □ 12.4

```
ps <- createPs(cohortMethodData = cmData,  
                population = studyPop)  
plotPs(ps, showCountsLabel = TRUE, showAucLabel = TRUE)
```



□ □ 12.5

```
showMaxLabel = TRUE,  
beforeLabel = "  ",  
afterLabel = "  ")
```



0.1 x > 0.3

□ □ 12.6

PS□□□□ Cox □□□□□□□□□□□□□□□□□□

```
##      cox
##      TRUE
##      FALSE
##      FALSE
##      OK
##
##          .95    .95
##      1.13211   0.92132   1.40008  0.12409   0.1068
```

E.9

□ □ 13.1

getP1pData

```
## plpData
##
##           ID -1
##      ID 3
##
##      2630
##
##      479      479
## 3
##
##      245
##
##      54079
```

□ □ 13.2

```
population <- createStudyPopulation(plpData = plpData,
                                     outcomeId = 3,
                                     washoutPeriod = 364,
                                     firstExposureOnly = FALSE,
                                     removeSubjectsWithPriorOutcome = TRUE,
                                     priorOutcomeLookback = 9999,
                                     riskWindowStart = 1,
                                     riskWindowEnd = 365,
                                     addExposureDaysToStart = FALSE,
                                     addExposureDaysToEnd = FALSE,
                                     minTimeAtRisk = 364,
                                     requireTimeAtRisk = TRUE,
                                     includeAllOutcomes = TRUE)

nrow(population)
```

```
## [1] 2578
```

□ □ 13.3

LASSO runPlp

```
lassoModel <- setLassoLogisticRegression(seed = 0)

lassoResults <- runPlp(population = population,
                        plpData = plpData,
                        modelSettings = lassoModel,
                        testSplit = 'person',
                        testFraction = 0.25,
                        nfold = 2,
                        splitSeed = 0)
```

LASSO

Shiny

```
viewPlp(lassoResults)
```

E 18 AUC 0.645

The screenshot shows a Shiny application interface titled "PatientLevelPrediction Explorer". The top navigation bar includes tabs for "Internal Validation" and "External Validation". Below the navigation bar, a horizontal menu bar contains links for "Evaluation Summary", "Characterization", "ROC", "Calibration", "Demographics", "Preference", "Box Plot", and "Settings". The main content area is titled "Evaluation Summary". It features a table with the following data:

Metric	test	train
1 AUC	0.645	0.7112
2 AUC_lb95ci	0.589	0.6815
3 AUC_ub95ci	0.700	0.7409
4 AUPRC	0.286	0.3615
5 BrierScaled	0.062	0.0860
6 BrierScore	0.144	0.1382

At the top left of the table, there is a dropdown menu labeled "Show 25 entries". At the top right, there is a search bar labeled "Search: []".

Figure E.18: □□□□□□□Shiny□□□□

E.10

□□ 15.1

ACHILLES□□□□□□□:

```
library(ACHILLES)
result <- achilles(connectionDetails,
                      cdmDatabaseSchema = "main",
                      resultsDatabaseSchema = "main",
                      sourceName = "Eunomia",
                      cdmVersion = "5.3.0")
```

□□ 15.2

□□□□□□□□□□□□□□□□□□□□:

```
DataQualityDashboard::executeDqChecks(
  connectionDetails,
  cdmDatabaseSchema = "main",
  resultsDatabaseSchema = "main",
  cdmSourceName = "Eunomia",
  outputFolder = "C:/dataQualityExample")
```

□□ 15.3

□□□□□□□□□□□□□□□□□□□□:

```
DataQualityDashboard::viewDqDashboard(  
  "C:/dataQualityExample/Eunomia/results_Eunomia.json")
```

E.11

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