

## **Population-level Estimation**

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### Exercise: the Example from the Book of OHDSI

### book.ohdsi.org

- Chapter 12. Population-Level Estimation
- 12.6 Designing a Hypertension Study

### 12.6 Designing a Hypertension Study

# The Book of OHDSE: Korea CREATION CREATION

#### 12.6.1 Problem Definition

ACE inhibitors (ACEi) are widely used in patients with hypertension or ischemic heart disease, especially those with other comorbidities such as congestive heart failure, diabetes mellitus, or chronic kidney disease. (Zaman, Oparil, and Calhoun 2002) Angioedema, a serious and sometimes life-threatening adverse event that usually manifests as swelling of the lips, tongue, mouth, larynx, pharynx, or periorbital

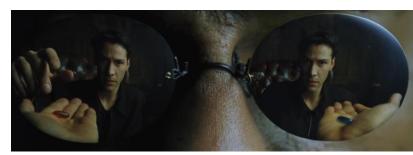


## A pop culture mash-up to explain counterfactual reasoning...





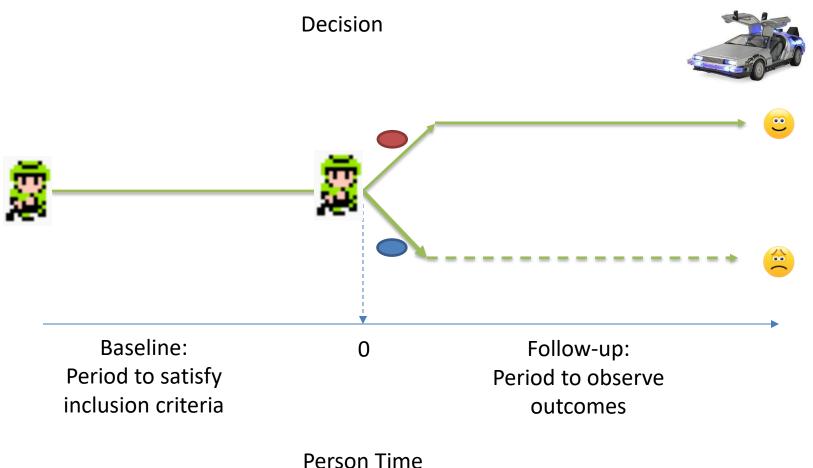






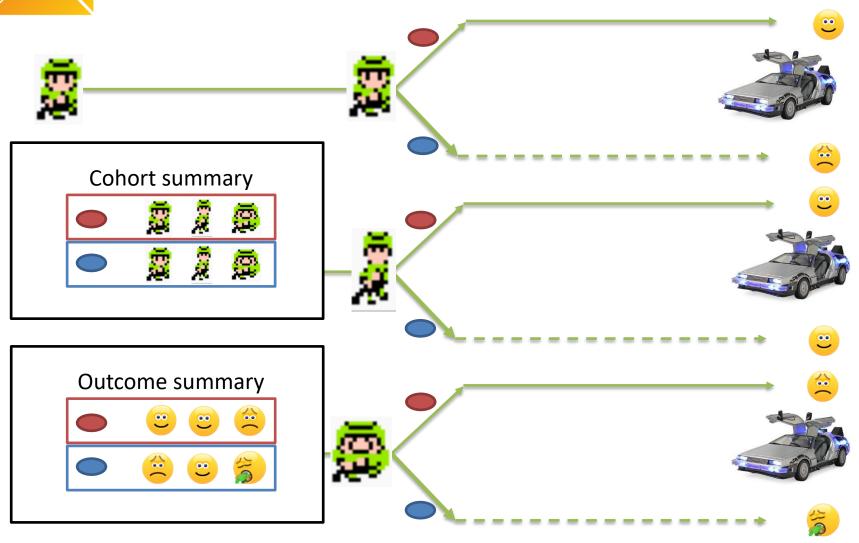


### Counterfactual reasoning for one person





### Counterfactual reasoning for a population





### OHDSI's definition of 'cohort'

## Cohort = a set of persons who satisfy one or more inclusion criteria for a duration of time

Objective consequences based on this cohort definition:

- One person may belong to multiple cohorts
- One person may belong to the same cohort at multiple different time periods
- One person may not belong to the same cohort multiple times during the same period of time
- One cohort may have zero or more members
- A codeset is NOT a cohort...

...logic for how to use the codeset in a criteria is required



## Process flow for formally defining a cohort in ATLAS

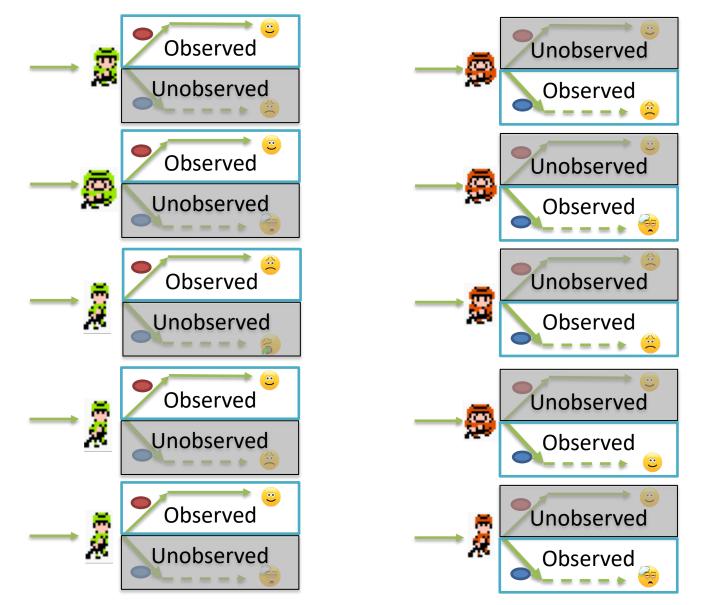
- Cohort entry criteria
  - Initial events
    - Events are recorded time-stamped observations for the persons, such as drug exposures, conditions, procedures, measurements and visits.
    - All events have a start date and end date, though some events may have a start date and end date with the same value (such as procedures or measurements).
  - Initial event inclusion criteria
  - Additional qualifying inclusion criteria
    - The qualifying cohort will be defined as all persons who have an initial event, satisfy the initial event inclusion criteria, and fulfill all additional qualifying inclusion criteria.
    - Each qualifying inclusion criteria will be evaluated to determine the impact of the criteria on the attrition of persons from the initial cohort.
- Cohort exit criteria

Initial cohort

Qualifying cohort



## An observational comparative cohort design to approximate counterfactual outcomes





## Propensity score introduction

- Propensity score = probability of belonging to the target cohort vs. the comparator cohort, given the baseline covariates
- e(x) = Pr(Z=1|x)
  - Z is treatment assignment
  - x is a set of all covariates at the time of treatment assignment
- Propensity score can be used as a 'balancing score': if the two cohorts have similar propensity score distribution, then the distribution of covariates should be the similar (need to perform diagnostic to check)

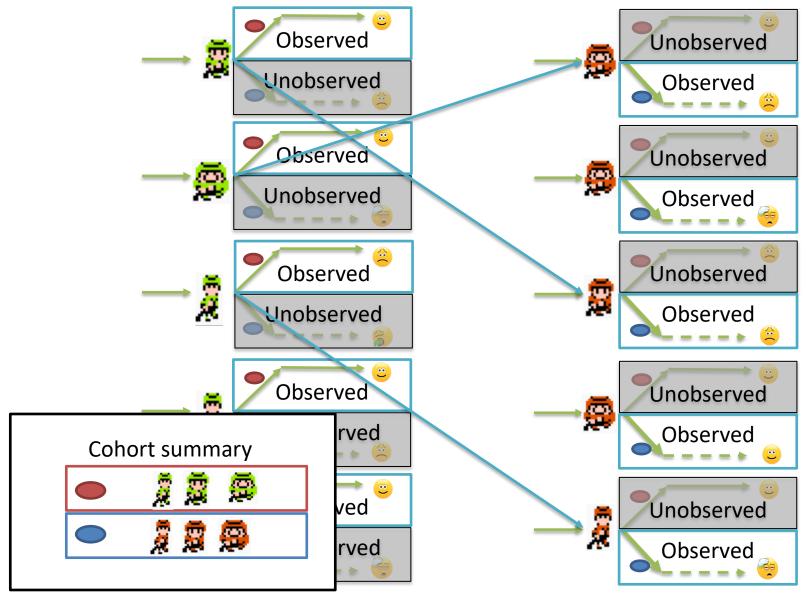


## Methods for confounding adjustment using a propensity score

Regression adjustment	The PS is used as a covariable in an outcome regression model to adjust		
	Not generally recommended e		
	relationship between propensity score and outcome is correctly specified.		
Matching	The PS is used to match exposed subjects to unexposed subjects with similar values of the PS. This method assumes that within the matched sample, exposed and unexposed subjects have a similar distribution of baseline characteristics.		
Stratification	The PS is used to stratify subjects into (often quintiles or deciles) strata.  Treatment effects are estimated separately within each stratum and then combined into an overall estimate of treatment effect. This method assumes that within each stratum, exposed and unexposed subjects have a similar distribution of baseline characteristics.		
Inverse Probability Weighting  The PS is used to create weights based on the inverse probability defined as: E*/PS + (1-E)/(1-PS). This assumes that characteristics are similar in the exposed and unexposed group.			
	Fully implemented in OHDS		
* E: exposure	CohortMethod R package		

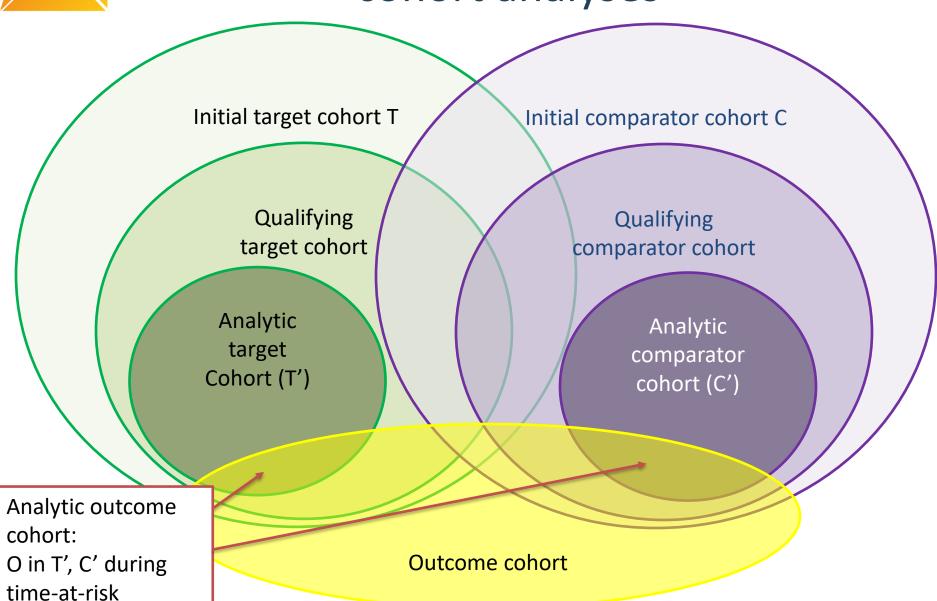


## Matching as a strategy to adjust for baseline covariate imbalance





## Cohort restriction in comparative cohort analyses





## The choice of the outcome model defines your research question

	Logistic regression	Poisson regression	Cox proportional hazards
How the outcome cohort is used	Binary classifier of presence/ absence of outcome during the fixed timeat-risk period	Count the number of occurrences of outcomes during time-at-risk	Compute time-to-event from time-at-risk start until earliest of first occurrence of outcome or time-at-risk end, and track the censoring event (outcome or no outcome)
'Risk' metric	Odds ratio	Rate ratio	Hazard ratio
Key model assumptions	Constant probability in fixed window	Outcomes follow Poisson distribution with constant risk	Proportionality – constant relative hazard



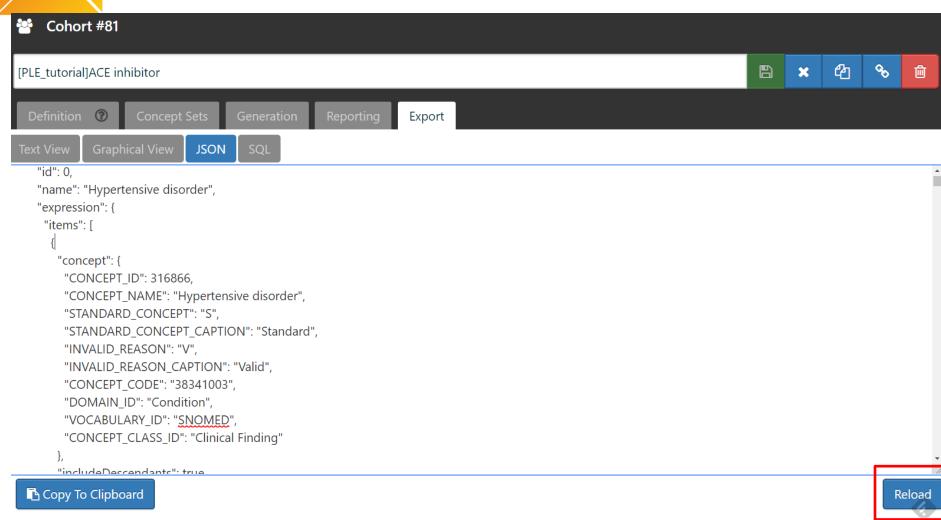
## Population-level Estimation <Exercise>

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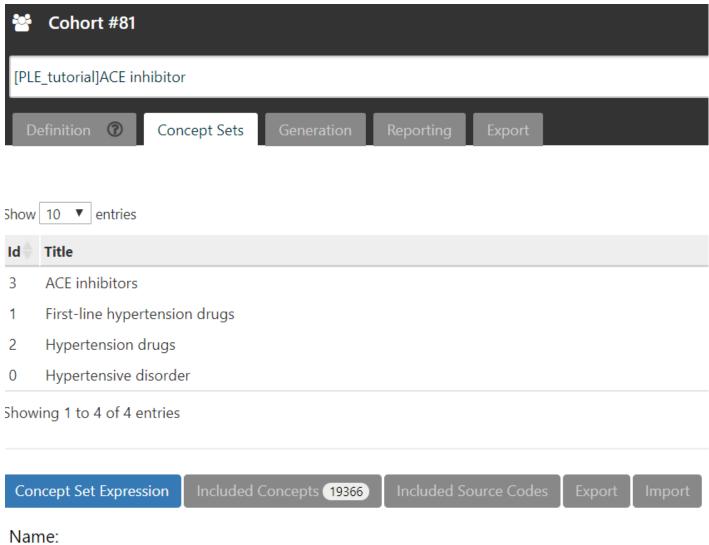


### **Load Cohort Definitions**

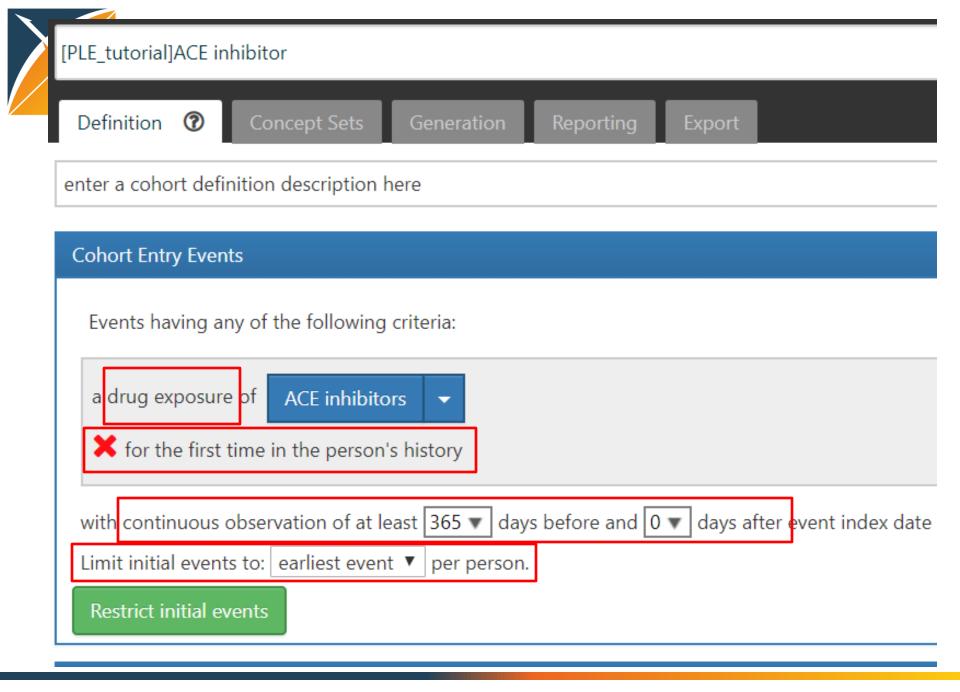


https://github.com/OHDSI/TheBookOfOhdsi/tree/master/extras/ CohortMethodAceiVsThz/inst/cohorts

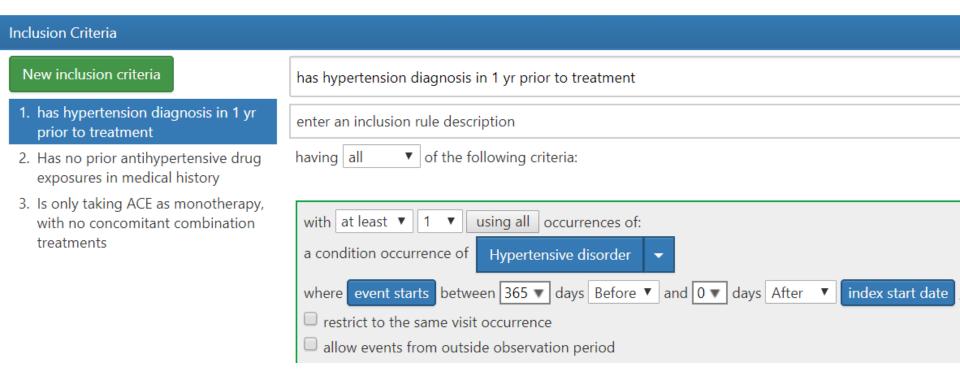




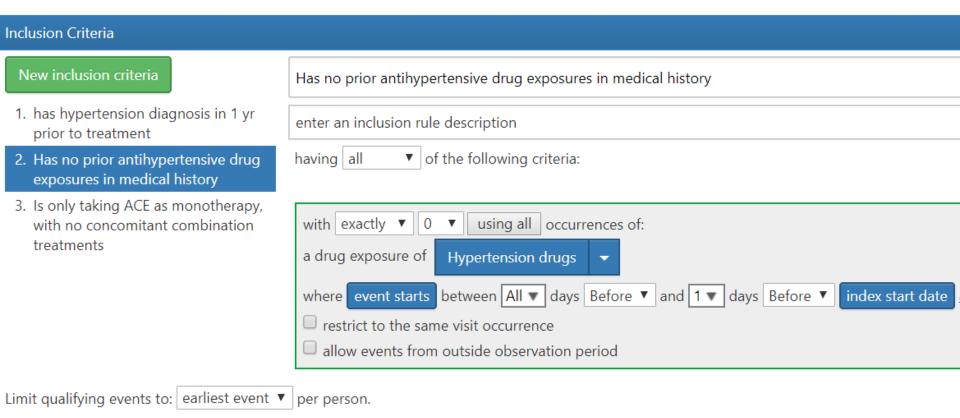
**ACE** inhibitors



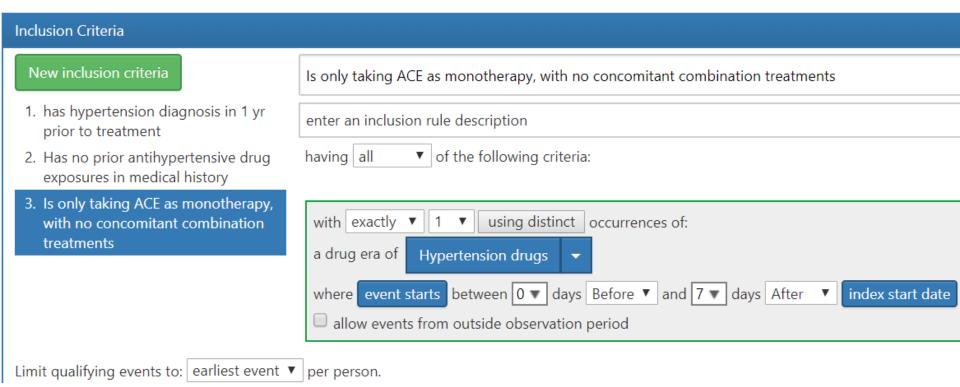














#### Cohort Exit

#### **Event Persistence:**

Event will persist until: end of a continuous drug exposure ▼

#### **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 conce 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 inference 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 30 🔻 days between exposure records when inferring the era of persistence exposure
- Surveillance window: add 0 ▼ 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.



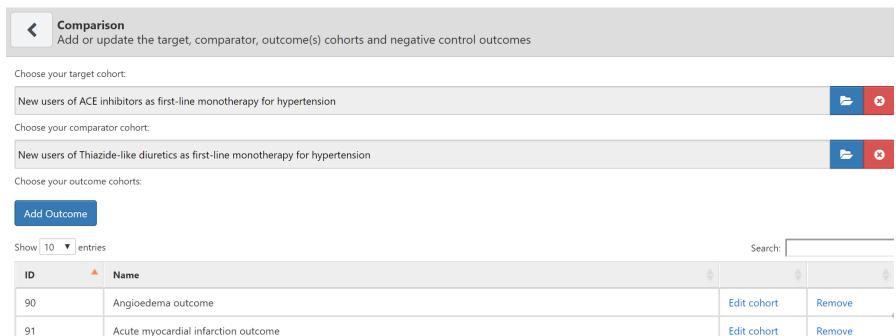
## Result from generation

Available CDM Sources					
	Source Name	Generation Status	People	Records	
<b>▶</b> Generate	CMSDESynPUF100k	COMPLETE	261	261	
<b>▶</b> Generate	CMSDESynPUF1k	COMPLETE	21	21	
<b>▶</b> Generate	CMSDESynPUF23m	COMPLETE	5,124	5,124	

	Summary Statistics:	Match Rate 1.62%	Matches 5,124		1 <b>Events</b> 16,688		Attrition Visualization	Switch to intersect view
Inclusion Rule				N	% Remain	% Diff		
1. has hypertension diagnosis in 1 yr prior to treatment				25,979	8.20%	91.80%		
2. Has no prior antihypertensive drug exposures in medical history				5,943	1.88%	6.33%		
3. Is only taking ACE	as monotherapy, with no concomitant comb	bination treatment	is .	5,124	1.62%	0.26%		



## Comparison setting





## Add negative controls and Concepts to exclude

Showing 1 to 2 of 2 entries

Choose your negative control outcomes:

Negative controls for ACEi and THZ

Covariate selection

**Please note:** If you would like to include/exclude covariates based on descendant conce descendants, define your concept sets utilizing **the ancestor concepts only**.

What concepts do you want to include in baseline covariates in the propensity score model? (Le

What concepts do you want to exclude from baseline covariates in the propensity score model?

Concepts to exclude for ACEi and THZ



No ▼

## **Analysis Settings**

Should only the first exposure per subject be included?
No ▼
Remove subjects that are in both the target and comparator cohort?
Remove All ▼
Restrict the analysis to the period when both exposures are observed?
No ▼
The mininum required continuous observation time prior to index date for a person to be included in the cohort.
0
If either the target or the comparator cohort is larger than this number it will be sampled to this size. (0 for this value
0 🔻
Remove subjects that have the outcome prior to the risk window start?
Yes ▼
How many days should we look back when identifying prior outcomes?
99999 🔻
If a subject is in multiple cohorts, should time-at-risk be censored when the new time-at-risk start to prevent overlap



## Changing Time At Risk

#### Time At Risk

Define the time-at-risk window start, relative to target/comparator cohort entry:

0 ▼ days from cohort start date ▼

Define the time-at-risk window end:

365 ▼ days from cohort end date ▼

The minimum number of days at risk?

0 ▼



## **Covariate Settings**

Covariate Settings
Using OHDSI covariates for propensity score model. Click to view details)  What concepts do you want to <b>include</b> in baseline covariates in the propensity score
Should descendant concepts be added to the list of included concepts?  No ▼
What concepts do you want to <b>exclude</b> in baseline covariates in the propensity so
Should descendant concepts be added to the list of excluded concepts?  Yes ▼
A comma delimited list of covariate IDs that should be restricted to:



#### A Propensity Score Adjustment

How do you want to trim your cohorts based on the propensity score distribution?

None ▼

Do you want to perform matching or stratification?

Match on propensity score ▼

What is the maximum number of persons in the comparator arm to be matched to each person in the target arm within the defined caliper? (0 = means no maximum

100 ▼

What is the caliper for matching:

0.2

What is the caliper scale:

Standardized Logit ▼

What is the maximum number of people to include in the propensity score model when fitting? Setting this number to 0 means no down-sampling will be applied:

250000 ▼

Test each covariate for correlation with the target assignment? If any covariate has an unusually high correlation (either positive or negative), this will throw an error.

Yes ▼

If an error occurs, should the function stop? Else, the two cohorts will be assumed to be perfectly separable.

Yes ▼



## Outcome Model Settings

Choice	Value
Model	Cox proportional hazards model using variable-ratio matching.

#### **C**Outcome Model Settings

Specify the statistical model used to estimate the risk of outcome between target and comparator cohorts:

Cox proportional hazards ▼

Should the regression be conditioned on the strata defined in the population object (e.g. by matching or stratifying on propensity scores)?

Yes ▼

Whether to use the covariate matrix in the cohortMethodDataObject in the outcome model.

No ▼

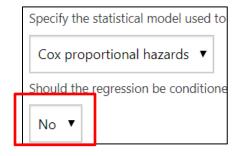
Use inverse probability of treatment weighting?

No ▼

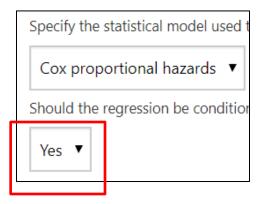


## Recommended statistical model settings

 1:1 PS matching and unconditioned Cox regression



 Variable-ratio (1:100) PS matching and conditioned Cox regression



 PS stratification and conditioned Cox regression



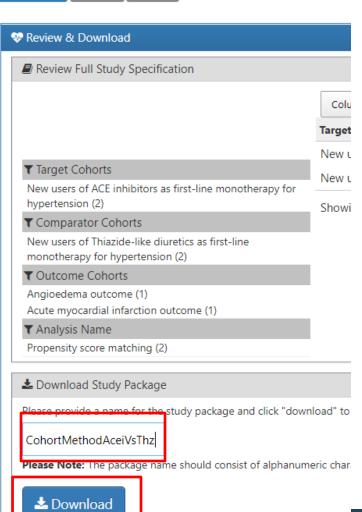
## Download the package



Download

Import

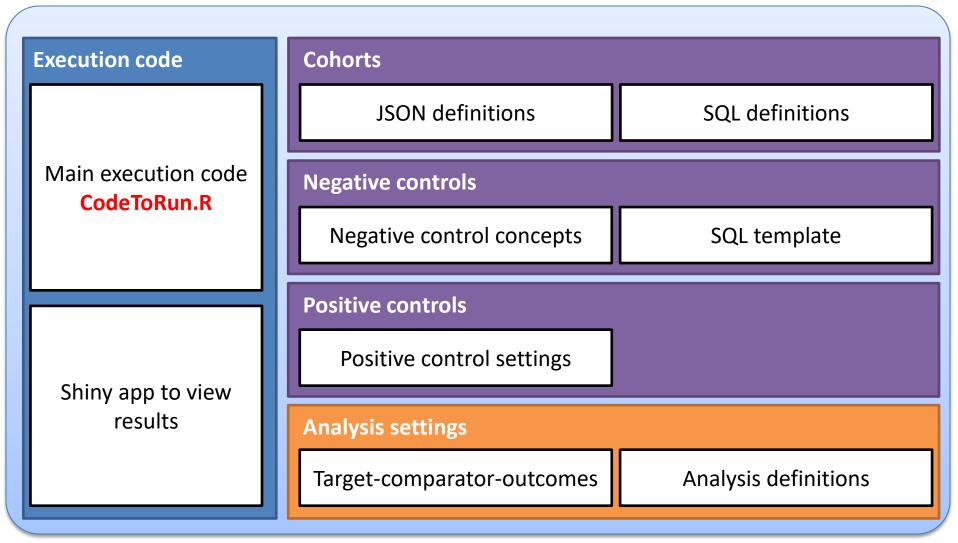
xport



 Name the package and click 'Download' in the Utilities tab



## Anatomy of the study package





## Running package

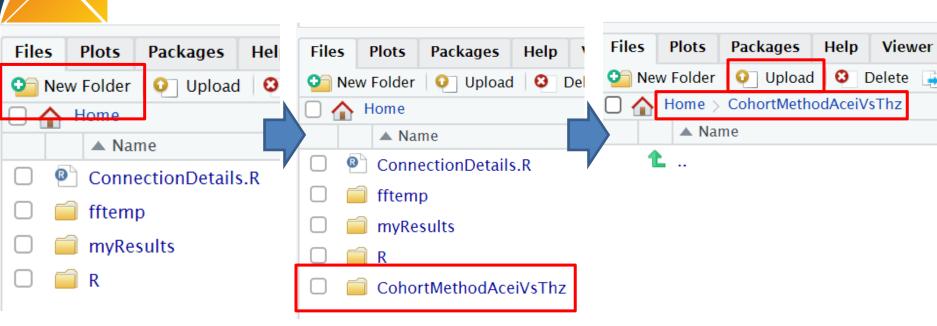
Source: readme.md

3. Once installed, you can execute the study by modifying and using the following code:

```
library (Graham)
# Optional: specify where the temporary files (used by the ff package) will be created:
options(fftempdir = "c:/FFtemp")
# Maximum number of cores to be used:
maxCores <- parallel::detectCores()
# Minimum cell count when exporting data:
minCellCount <- 5
# The folder where the study intermediate and result files will be written:
outputFolder <- "c:/Graham"
# Details for connecting to the server:
# See ?DatabaseConnector::createConnectionDetails for help
connectionDetails <- DatabaseConnector::createConnectionDetails(dbms = "postgresgl",
                                server = "some.server.com/ohdsi",
                                user = "joe",
                                password = "secret")
# The name of the database schema where the CDM data can be found:
cdmDatabaseSchema <- "cdm synpuf"
```

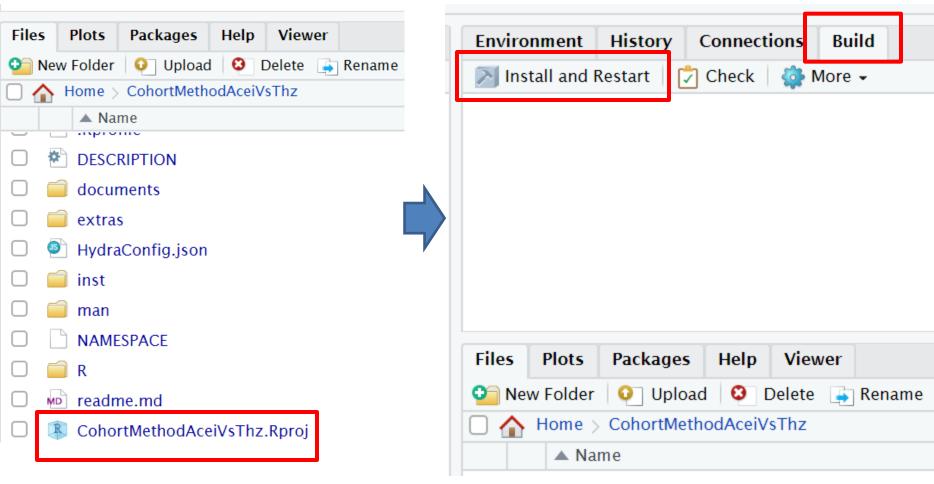


## Upload the package



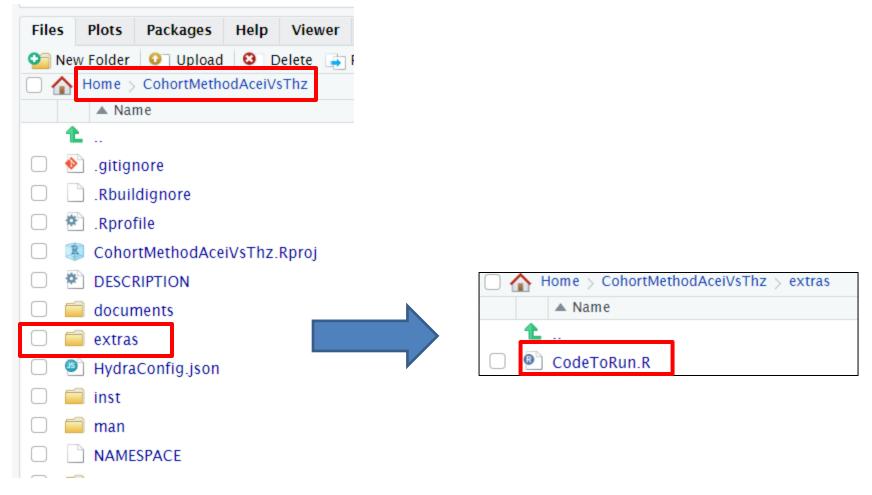


## Install R package





### CodeToRun.R





### Execute the Code

```
25
    cohortTable <- "mschuemi skeleton"</pre>
26
27
    #Connection string for the OMOP database on Redshift
    connectionDetails <- DatabaseConnector::createConnectionDetails(dbms = "redshift",</pre>
28
                                                                       server = "ajou-ohdsi-data
29
30
                                                                       user = "master",
31
                                                                       password = "Ajoumed01",
32
                                                                       port = "5439")
33
34
    options(fftempdir = "~/fftemp")
35
    outputFolder <- "myResults"
36
37
    cdmDatabaseSchema <- "CMSDESynPUF100k"
    cohortDatabaseSchema <- "CMSDESynPUF100kresults"</pre>
38
39
40
   # Some meta-information that will be used by the export function:
    databaseId <- "Synpuf"
41
    databaseName <- "Medicare Claims Synthetic Public Use Files (SynPUFs)"
42
    databaseDescription <- "Medicare Claims Synthetic Public Use Files (SynPUFs) were created
43
44
45
    # For Oracle: define a schema that can be used to emulate temp tables:
46
    oracleTempSchema <- NULL
47
```



### Execute the Code

```
execute(connectionDetails = connectionDetails,
        cdmDatabaseSchema = cdmDatabaseSchema,
        cohortDatabaseSchema = cohortDatabaseSchema,
        cohortTable = cohortTable,
        oracleTempSchema = oracleTempSchema,
        outputFolder = outputFolder,
        databaseId = databaseId,
        databaseName = databaseName,
        databaseDescription = databaseDescription,
        createCohorts = TRUE.
        synthesizePositiveControls = TRUE,
        runAnalyses = TRUE,
        runDiagnostics = TRUE,
        packageResults = TRUE,
        maxCores = maxCores)
resultsZipFile <- file.path(outputFolder, "export", paste0("Results", databaseId, ".zip"))
dataFolder <- file.path(outputFolder, "shinyData")</pre>
prepareForEvidenceExplorer(resultsZipFile = resultsZipFile, dataFolder = dataFolder)
```



## Check the result by ShinyViewer

69

launchEvidenceExplorer(dataFolder = dataFolder, blind = FALSE, launch.browser = FALSE)

blind = FALSE, launch.browser = FALSE

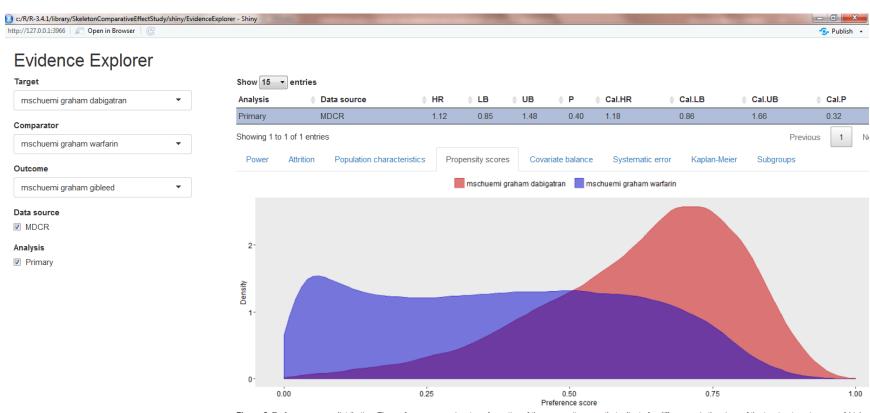


Figure 2. Preference score distribution. The preference score is a transformation of the propensity score that adjusts for differences in the sizes of the two treatment groups. A higher overlap indicates subjects in the two groups were more similar in terms of their predicted probability of receiving one treatment over the other.

♣ Download plot



## Concluding remarks

- CohortMethod package + R offer large flexibility
- 80% of studies are 'cookie-cutter' design, supported by ATLAS
- For remaining 20%, will need to modify code generated by ATLAS



## Thank you

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