



Large-scale Evidence Generation and Evaluation across a Network of Databases (**LEGEND**) Initiative



LEGEND: Principles and practice

Anna Ostropolets
on behalf of the LEGEND initiative



Towards Reliable Evidence ...

Some current practices across the broader research community

Examining one target-comparator pair at a time

Not using appropriate methods to control for bias

Modify the design until significant results are found



Large-scale evidence generation across a network of databases (LEGEND)

- * Pre-specified fixed design and dissemination of the results regardless of the estimates (avoid publication bias)

- * Systematic process across all research questions

- * Large-scale: looking at thousands of target-comparator pairs at a time

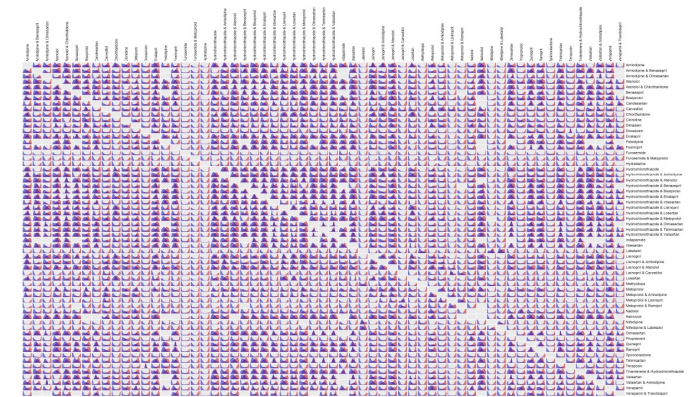
- * Use of best practices: LSPS, extensive diagnostics, negative and positive controls



Some of the LEGEND Principles Step-by-Step

OHDSI Large-Scale Evidence Generation and Evaluation in a Network of Databases (LEGEND): Study of the Effects of Treatments for Hypertension

Version: 0.1



Select multiple target and comparator cohorts,
for example, all drug in a drug class

- 39 mono-drugs, 13 mono-classes
- 58 duo-drugs, 32 duo-classes
- 10,278 comparisons

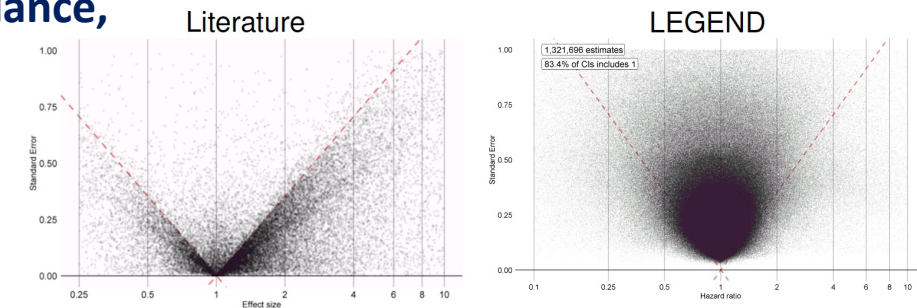
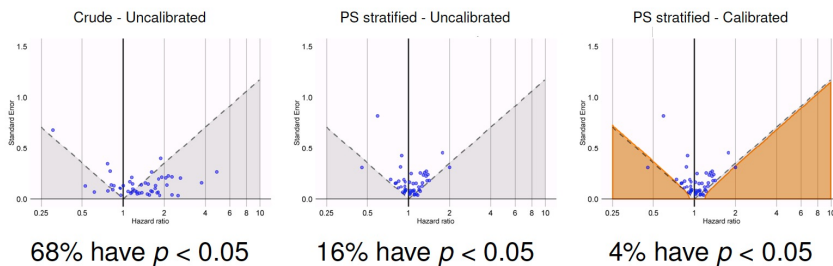
	Theoretical	Observed (n > 2,500)
Single ingredients	58	39
Single ingredient comparisons	$58 * 57 = 3,306$	1,296
Single drug classes	15	13
Single class comparisons	$15 * 14 = 210$	156
Dual ingredients	$58 * 57 / 2 = 1,653$	58
Single vs duo drug comparisons	$58 * 1,653 = 95,874$	3,810
Dual classes	$15 * 14 / 2 = 105$	32
Single vs duo class comparisons	$15 * 105 = 1,575$	832
Duo vs duo drug comparisons	$1,653 * 1,652 = 2,730,756$	2,784
Duo vs duo class comparisons	$105 * 104 = 10,920$	992
...
Total comparisons	2,843,250	10,278

Carefully design the study, including
sensitivity analyses

Run on multiple databases

Spend a lot of time on diagnostics:
propensity score balance, covariate balance,
calibration plots etc.

Publish all results





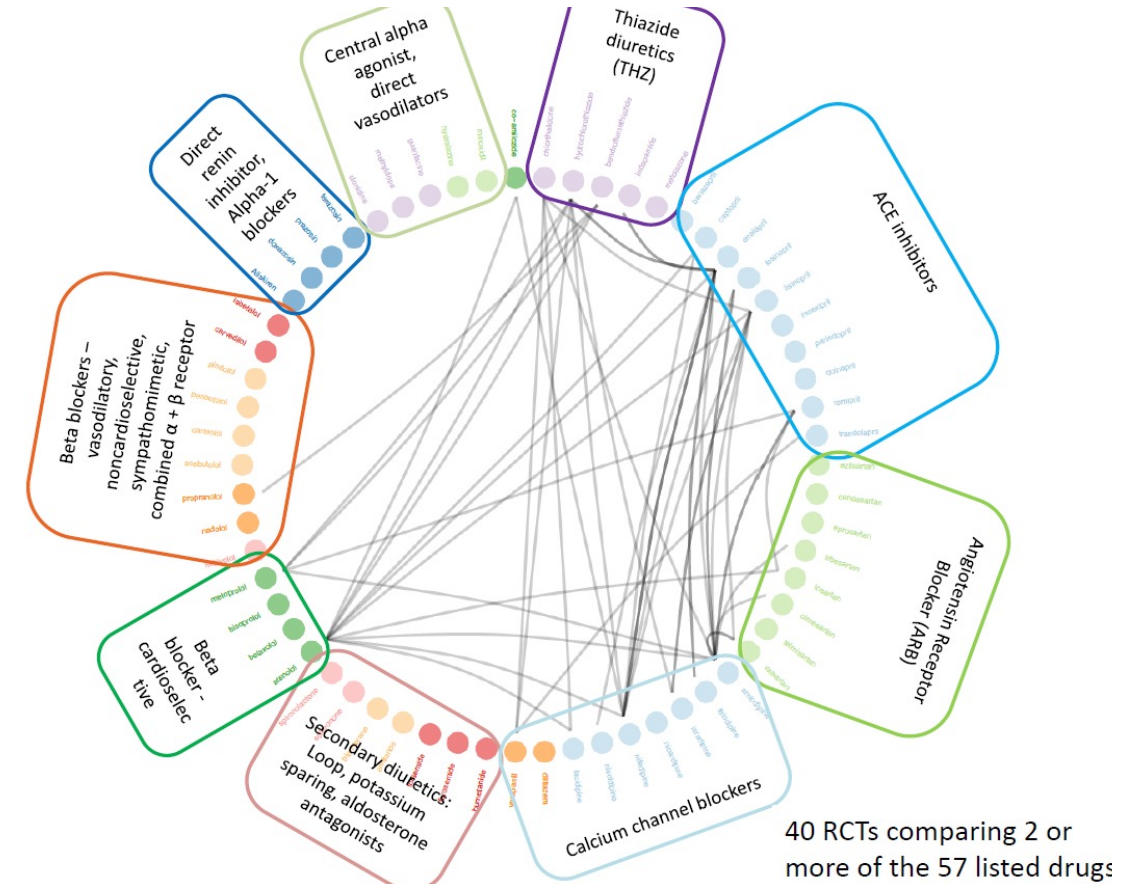
LEGEND-HTN: First-line antihypertensive treatments

RuiJun (Ray) Chen
on behalf of the LEGEND initiative



LEGEND-HTN Motivation

- Why?
 - New 2017, 2018 guidelines
 - Multiple recommended 1st-line drug classes
 - Thiazides, ACE inhibitors, ARBs, CCBs, B-blockers (ESC/ESH only)
 - Older RCTs, few direct comparisons
- Can we do better?





LEGEND-HTN Prior Publications

Main manuscript: class-level comparison, 55 outcomes, 9 databases

Suchard, et al. Published 2019 in *The Lancet*

Articles

Comprehensive comparative effectiveness and safety of first-line antihypertensive drug classes: a systematic, multinational, large-scale analysis



Marc A Suchard, Martijn J Schuemie, Harlan M Krumholz, Seng Chan You, Ruijun Chen, Nicole Pratt, Christian G Reich, Jon Duke, David Madigan, George Hripcsak, Patrick B Ryan

Summary

Background Uncertainty remains about the optimal monotherapy for hypertension, with current guidelines recommending any primary agent among the first-line drug classes thiazide or thiazide-like diuretics, angiotensin-converting enzyme inhibitors, angiotensin receptor blockers, dihydropyridine calcium channel blockers, and non-dihydropyridine calcium channel blockers, in the absence of comorbid indications. Randomised trials have not further refined this choice.

Published Online
October 24, 2019
[https://doi.org/10.1016/S0140-6736\(19\)32317-7](https://doi.org/10.1016/S0140-6736(19)32317-7)



LEGEND-HTN Prior Publications

Chlorthalidone vs HCTZ

- Chlorthalidone recommended as preferred thiazide diuretic
 - Longer half life
 - Used in trials
- HCTZ is most commonly prescribed thiazide
- No direct head-to-head comparisons (RCT in progress)
- Hripcsak, et al.
Published 2020 in JAMA Internal Medicine

Research

JAMA Internal Medicine | [Original Investigation](#)

Comparison of Cardiovascular and Safety Outcomes of Chlorthalidone vs Hydrochlorothiazide to Treat Hypertension

George Hripcsak, MD, MS; Marc A. Suchard, MD, PhD; Steven Shea, MD; RuiJun Chen, MD; Seng Chan You, MD; Nicole Pratt, PhD; David Madigan, PhD; Harlan M. Krumholz, MD, SM; Patrick B. Ryan, PhD; Martijn J. Schuemie, PhD

IMPORTANCE Chlorthalidone is currently recommended as the preferred thiazide diuretic to treat hypertension, but no trials have directly compared risks and benefits.

OBJECTIVE To compare the effectiveness and safety of chlorthalidone and hydrochlorothiazide as first-line therapies for hypertension in real-world practice.

[+ Supplemental content](#)

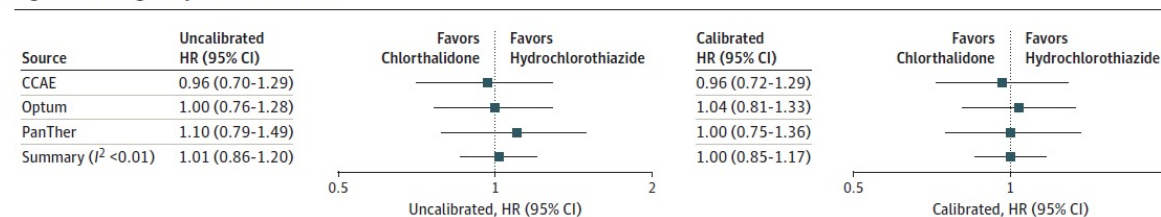


LEGEND-HTN Prior Publications

Chlorthalidone vs HCTZ

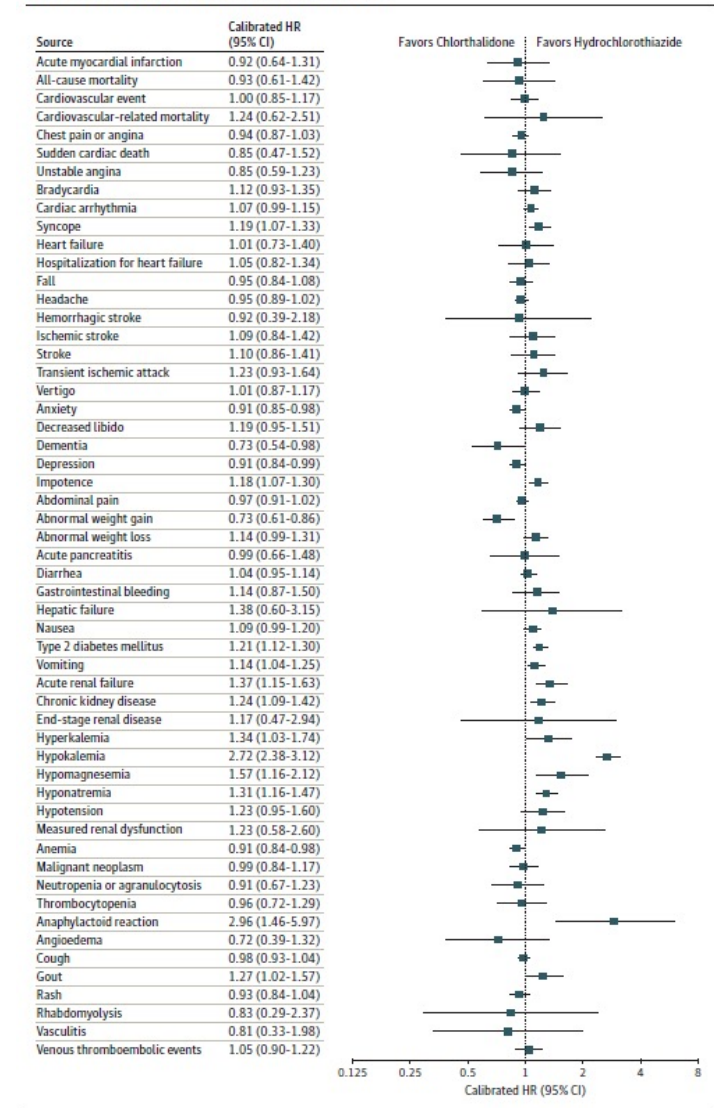
- 3 databases with >2500 exposures
 - 730K patients
- No significant difference in effectiveness
- HCTZ with fewer renal and electrolyte abnormalities

Figure 2. Homogeneity on Effectiveness



Hazard ratios (HRs) and forest plot of the 3 databases and the meta-analysis for chlorthalidone vs hydrochlorothiazide on the composite cardiovascular disease outcome. The 3 databases showed excellent agreement. CCAE indicates Commercial Claims and Encounters Database.

Figure 3. Forest Plot of Safety and Effectiveness Outcomes





LEGEND-HTN Prior Publications

Beta blockers

- ACC/AHA guidelines no longer recommend beta blockers
- Heterogeneous
 - 3rd generation beta blockers have greater vasodilatory effects
- Few direct comparisons vs atenolol or other classes
- You et al.
Published 2021 in Hypertension

Hypertension

BETA-BLOCKER THERAPY

Comprehensive Comparative Effectiveness and Safety of First-Line β -Blocker Monotherapy in Hypertensive Patients

A Large-Scale Multicenter Observational Study

Seng Chan You, Harlan M. Krumholz¹, Marc A. Suchard², Martijn J. Schuemie³, George Hripcsak, RuiJun Chen⁴, Steven Shea⁵, Jon Duke, Nicole Pratt, Christian G. Reich⁶, David Madigan⁷, Patrick B. Ryan, Rae Woong Park, Sungha Park⁸

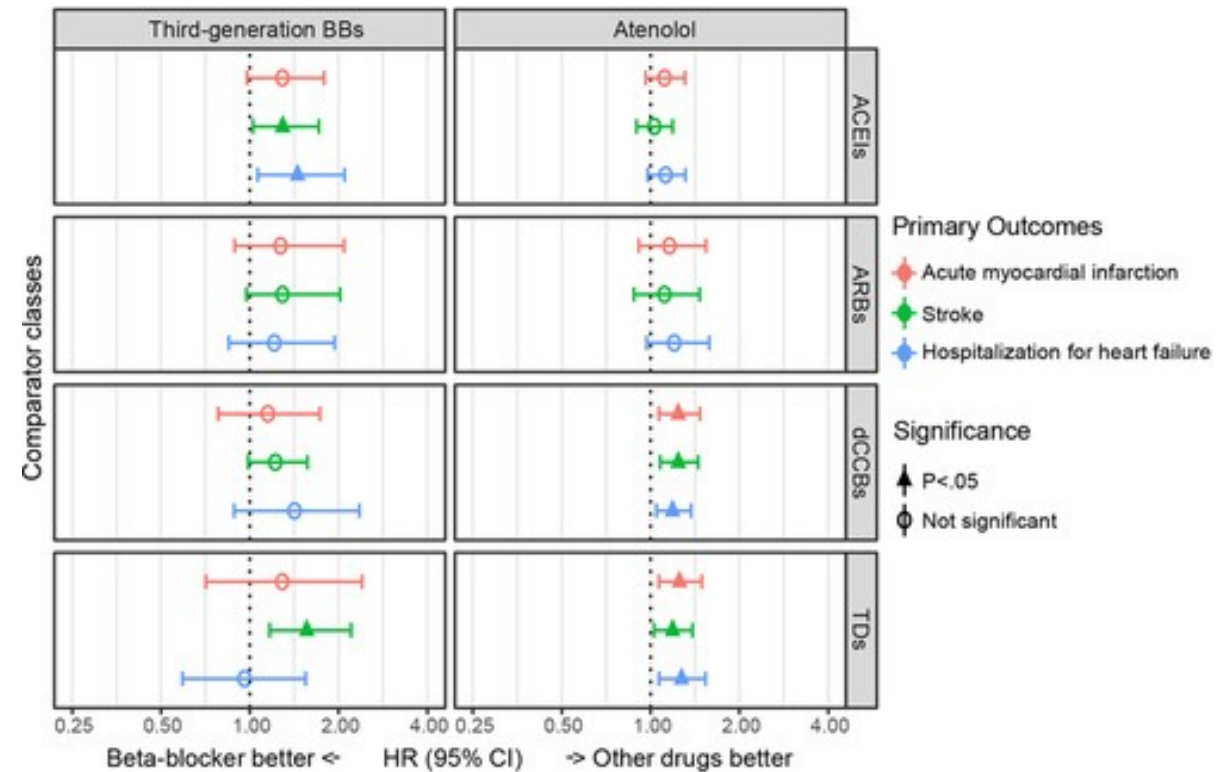
ABSTRACT: Evidence for the effectiveness and safety of the third-generation β -blockers other than atenolol in hypertension remains scarce. We assessed the effectiveness and safety of β -blockers as first-line treatment for hypertension using 3 databases in the United States: 2 administrative claims databases and 1 electronic health record–based database from 2001 to 2018. In each database, comparative effectiveness of β -blockers for the risks of acute myocardial infarction, stroke, and hospitalization for heart failure was assessed, using large-scale propensity adjustment and empirical calibration. Estimates were combined across databases using random-effects meta-analyses. Overall, 118 133 and 267 891 patients initiated third-generation β -blockers (carvedilol and nebivolol) or atenolol, respectively. The pooled hazard ratios (HRs) of



LEGEND-HTN Prior Publications

Beta blockers

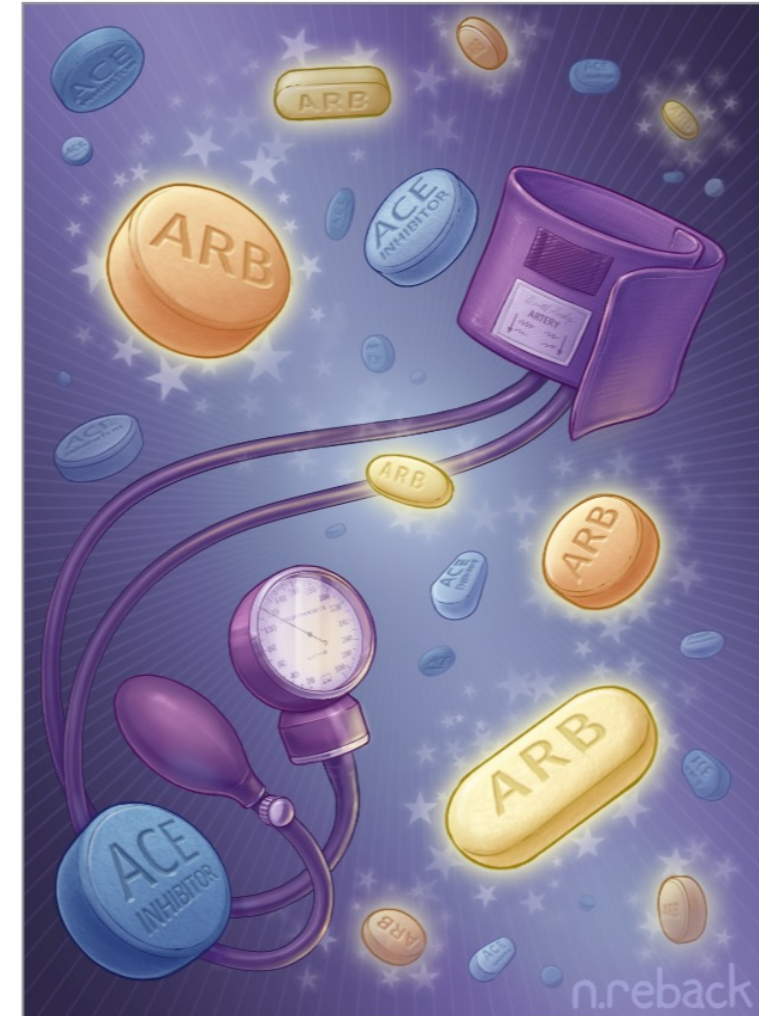
- 3 databases with >2500 exposures
 - 380K patients
- No significant difference in effectiveness between 3rd generation BBs and atenolol
- Higher risk for stroke with 3rd generation BBs vs ACE-I's or thiazide diuretics





LEGEND-HTN ACE Inhibitors vs ARBs

- Equally recommended 1st-line therapies
- Act along the same physiologic pathway
- Despite some known side effects, ACEs are much more commonly used than ARBs
- Few existing head-to-head studies





LEGEND-HTN ACE Inhibitors vs ARBs

- 8 databases with >2500 exposures
- 2.3M patients initiating treatment with ACE-I
- 674K patients initiating with ARBs
- No significant difference in primary outcomes

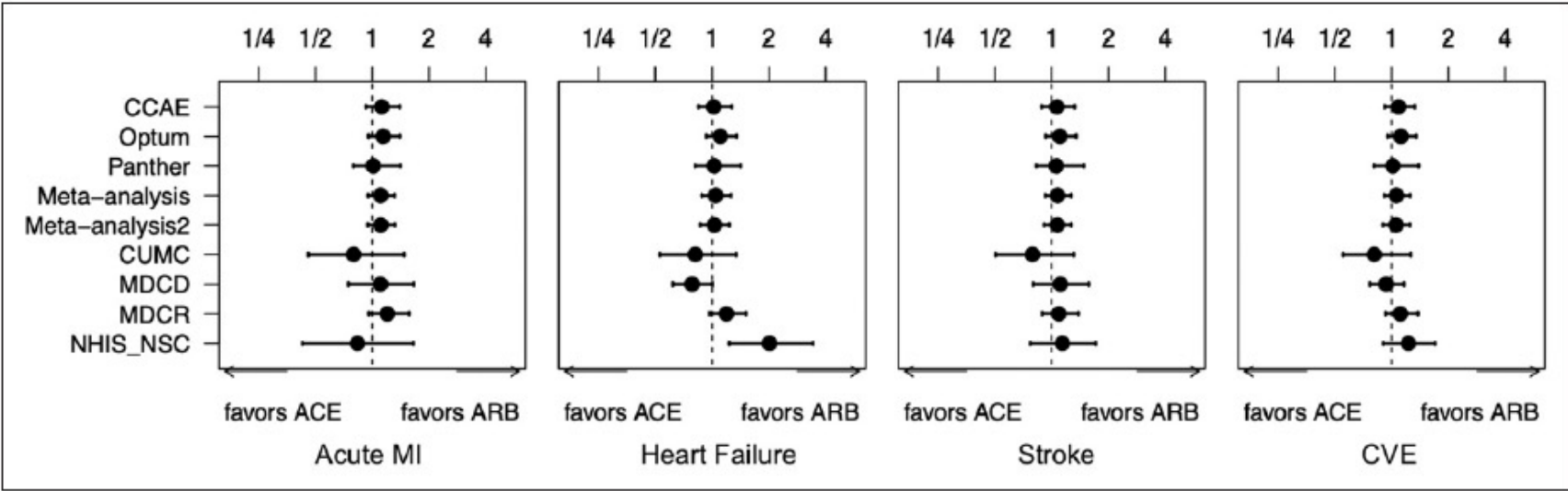
Hypertension

ANTIHYPERTENSIVE TREATMENT

Comparative First-Line Effectiveness and Safety of ACE (Angiotensin-Converting Enzyme) Inhibitors and Angiotensin Receptor Blockers

A Multinational Cohort Study

RuiJun Chen¹, Marc A. Suchard², Harlan M. Krumholz³, Martijn J. Schuemie⁴, Steven Shea⁵, Jon Duke, Nicole Pratt, Christian G. Reich⁶, David Madigan⁷, Seng Chan You, Patrick B. Ryan, George Hripcsak⁸





LEGEND-HTN ACE Inhibitors vs ARBs

- 51 secondary and safety outcomes
- ARBs with better safety profile
 - Lower risk of cough, angioedema, pancreatitis, and GI bleeding
 - Cough and angioedema remain significant even with Bonferroni correction

Outcome	HR (95% CI)	P value	Calibrated HR (95% CI)	Calibrated P value
Abdominal pain	1.00 (0.96–1.03)	0.87	1.01 (0.88–1.19)	0.87
Abnormal weight gain	0.82 (0.79–0.86)	<0.01	0.84 (0.74–0.98)	0.04
Abnormal weight loss	1.18 (1.11–1.25)	<0.01	1.18 (1.01–1.41)	0.04
Acute pancreatitis	1.32 (1.09–1.60)	<0.01	1.32 (1.04–1.70)	0.02
Acute renal failure	1.13 (1.08–1.18)	<0.01	1.14 (0.98–1.35)	0.10
Anaphylactoid reaction	1.31 (1.00–1.72)	0.05	1.31 (0.98–1.79)	0.07
Anemia	0.96 (0.92–0.99)	0.02	0.97 (0.84–1.14)	0.76
Angioedema	3.53 (2.99–4.16)	<0.01	3.31 (2.55–4.51)	<0.01
Anxiety	0.98 (0.95–1.00)	0.03	0.99 (0.86–1.16)	0.91
Bradycardia	0.96 (0.86–1.08)	0.52	0.98 (0.82–1.18)	0.84
Cardiac arrhythmia	0.96 (0.91–1.02)	0.22	0.98 (0.84–1.15)	0.82
Chest pain or angina	0.99 (0.97–1.01)	0.23	1.00 (0.87–1.17)	0.92
Chronic kidney disease	1.00 (0.93–1.08)	0.98	1.01 (0.87–1.20)	0.84
Cough	1.32 (1.23–1.42)	<0.01	1.32 (1.11–1.59)	<0.01
Decreased libido	0.96 (0.90–1.03)	0.29	0.98 (0.84–1.16)	0.83
Dementia	1.12 (1.06–1.18)	<0.01	1.13 (0.97–1.34)	0.14
Depression	1.02 (0.99–1.05)	0.20	1.03 (0.90–1.21)	0.65
Diarrhea	1.06 (1.02–1.09)	<0.01	1.07 (0.92–1.25)	0.40
End stage renal disease	0.87 (0.62–1.20)	0.39	0.88 (0.63–1.25)	0.50
Fall	1.00 (0.93–1.08)	0.98	1.01 (0.87–1.20)	0.84
Gastrointestinal bleed	1.18 (1.11–1.25)	<0.01	1.18 (1.01–1.41)	0.04
Gout	1.00 (0.97–1.04)	0.83	1.02 (0.88–1.19)	0.81



LEGEND-HTN: Dual combination therapy for treatment escalation

Yuan Lu

on behalf of the LEGEND initiative



Study Objective

As an extension of the LEGEND-HTN initiative, we aim to conduct a large-scale observational study within the OHDSI collaborative community to characterize real-world utilization of dual antihypertensive combination therapies for treatment escalation among people with hypertension.



Twelve Exposure Cohorts

Cohort #	1st Drug	2nd Drug
1	ACEi/ARB	CCB
2	CCB	ACEi/ARB
3	ACEi/ARB	Diuretic
4	Diuretic	ACEi/ARB
5	ACEi/ARB	Beta-blocker
6	Beta-blocker	ACEi/ARB
7	CCB	Diuretic
8	Diuretic	CCB
9	CCB	Beta-blocker
10	Beta-blocker	CCB
11	Diuretic	Beta-blocker
12	Beta-blocker	Diuretic



Data Sources

Data Source	Data Type	Country/District	Time Period	No. of Patients
IQVIA LPD Australia	EHR	Australia	2006-2020	3,101,500
ePBRN SWSLHD 2019 Linked Dataset (ePBRN SWSLHD)	EHR	Australia	2012-2019	139,346
Ajou University School of Medicine (AUSOM)	EHR	Korea	1995-2019	3,109,677
Kyung Hee University Hospital (KHMC)	EHR	Korea	2008-2018	2,010,456
Khoo Teck Puat Hospital (KTPH)	EHR	Singapore	2010-2016	290,074
National University Hospital (NUH)	EHR	Singapore	2015-2018	750,270
China Jiangsu Province Hospital (CJSPH)	EHR	China	2005-2015	6,230,000
Taiwan Taipei Medical University Clinical Research Database (TMUCRD)	EHR	Taiwan	2004-2020	3,659,572
IQVIA US Ambulatory EMR	EHR	United States	2006-2020	78,526,000
IQVIA LPD France	EHR	France	1994-2020	18,118,000
IQVIA LPD Italy	EHR	Italy	2004-2020	2,209,600

Together, the committed data sources cover:
118 millions patients in 8 countries and districts



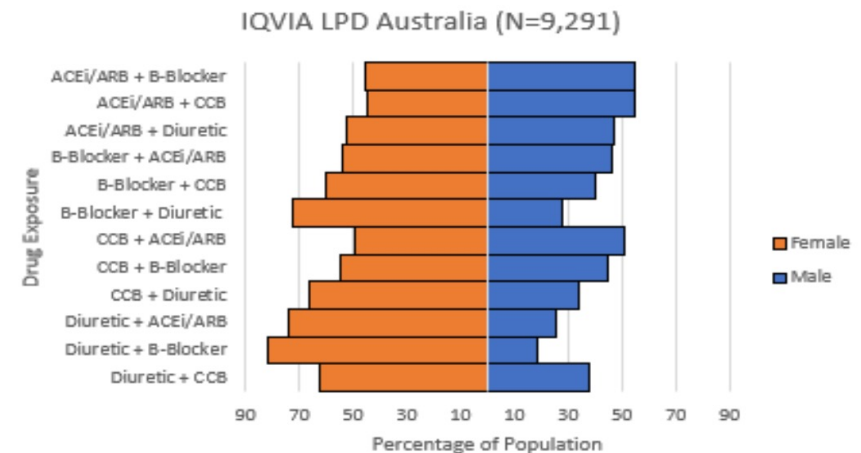
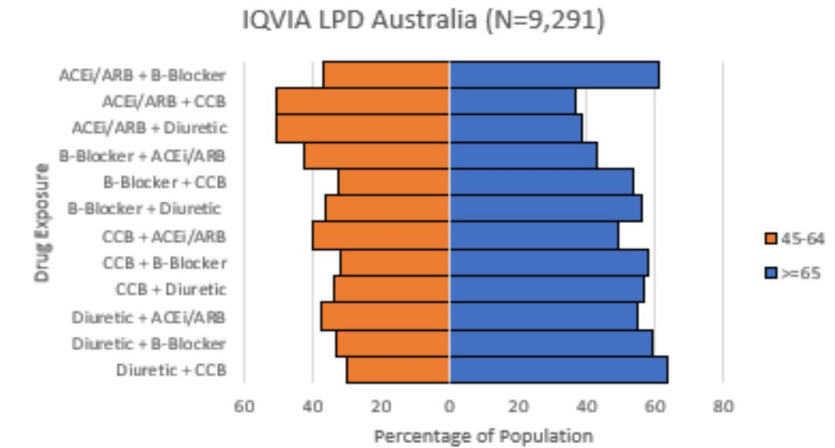
Patient Counts for 12 Exposure Cohorts

Cohort #	Dual combination	Data Sources										
		Australia		Korea		Singapore		China	Taiwan	France	Italy	United States
		Australia LPD	ePBRN SWSLHD	Ajou University	KHMC	KTPH	NUH	Jiangsu	TMUCRD	France LPD	Italy LPD	US AmbEMR
1	ACEi/ARB + Beta-blocker	1,184	268	392	49	105	144	46	1,464	11,236	11,844	110,579
2	ACEi/ARB + CCB	4,254	698	1,216	147	216	439	3,127	2,812	22,523	14,628	95,284
3	ACEi/ARB + Diuretic	2,066	508	474	12	16	31	111	8	22,399	16,988	123,940
4	Beta-blocker + ACEi/ARB	717	210	386	98	68	128	26	2,357	11,116	8,264	106,380
5	Beta-blocker + CCB	159	54	614	199	97	243	19	2,484	5,972	2,755	41,388
6	Beta-blocker + Diuretic	27	17	51	10	5	7	1	1	4,316	2,967	36,303
7	CCB + ACEi/ARB	1,339	246	1,487	191	191	133	3,312	5,015	15,749	5,841	54,297
8	CCB + Beta-blocker	190	41	814	217	120	101	34	2,518	3,866	2,475	30,593
9	CCB + Diuretic	74	28	259	15	11	6	78	4	1,660	1,103	21,108
10	Diuretic + ACEi/ARB	251	94	154	2	8	7	114	-	3,281	5,749	84,275
11	Diuretic + Beta-blocker	27	14	43	5	1	8	-	-	779	1,929	27,422
12	Diuretic + CCB	50	25	139	6	4	7	140	-	1,097	1,539	22,568



Cohort Characterization by Age and Sex

- Younger patients were more likely to be prescribed ACEi/ARB then a CCB or a diuretic compared with older patients.
- Women were more likely to be prescribed diuretics then an ACEi/ARB or a CCB compared with men.





Treatment Pathways

- Large variations treatment pathways across countries

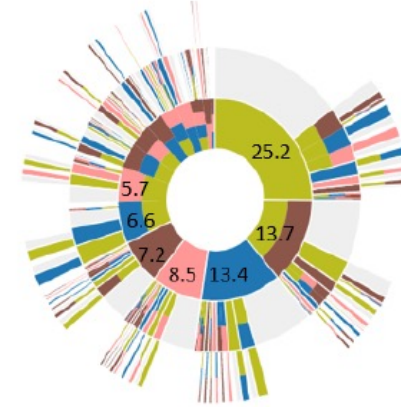
IQVIA US AmbEMR

Target Cohort

[APAC HTN] APAC overall population

- Target cohort count: 6,000,244
- Persons with pathways count: 5,166,727
- Persons with pathways portion: 86.1%

- [APAC HTN] ACEi/ARB use after hypertension diagnosis
- [APAC HTN] Diuretic use after hypertension diagnosis
- [APAC HTN] CCB use after hypertension diagnosis
- [APAC HTN] Beta-blocker use after hypertension diagnosis



IQVIA Italy LPD

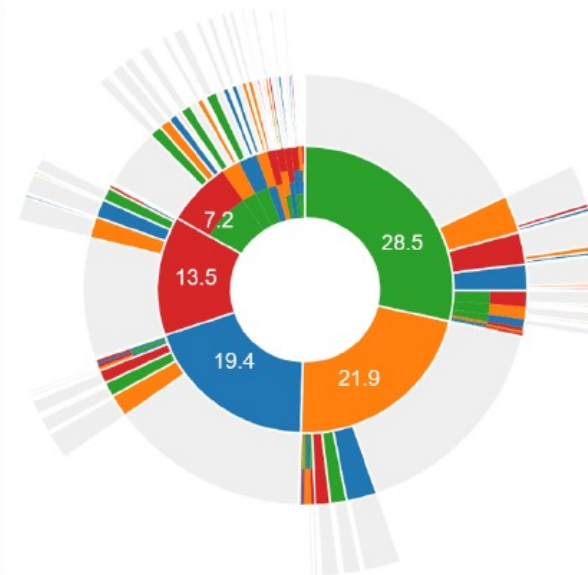
Target Cohort

[APAC HTN] APAC overall population

- Target cohort count: 535,157
- Persons with pathways count: 283,301
- Persons with pathways portion: 52.9%

Event Cohorts

- [APAC HTN] Beta-blocker use after hypertension diagnosis
- [APAC HTN] CCB use after hypertension diagnosis
- [APAC HTN] ACEi/ARB use after hypertension diagnosis
- [APAC HTN] Diuretic use after hypertension diagnosis





Main Findings and Lessons Learned

- Large variation in the transition between monotherapy and dual combination therapy for hypertension across countries and by demographic groups.



Main Findings and Lessons Learned

- Large variation in the transition between monotherapy and dual combination therapy for hypertension across countries and by demographic groups.
 - Future research is needed to identify what dual combinations work best for which patients.
-



Main Findings and Lessons Learned

- Large variation in the transition between monotherapy and dual combination therapy for hypertension across countries and by demographic groups.
- Future research is needed to identify what dual combinations work best for which patients.
- Using LEGEND principles can help mobilize collaboration with OHDSI data partners, but substantial effort was required to ensure data quality and alignment of methods across data sources.



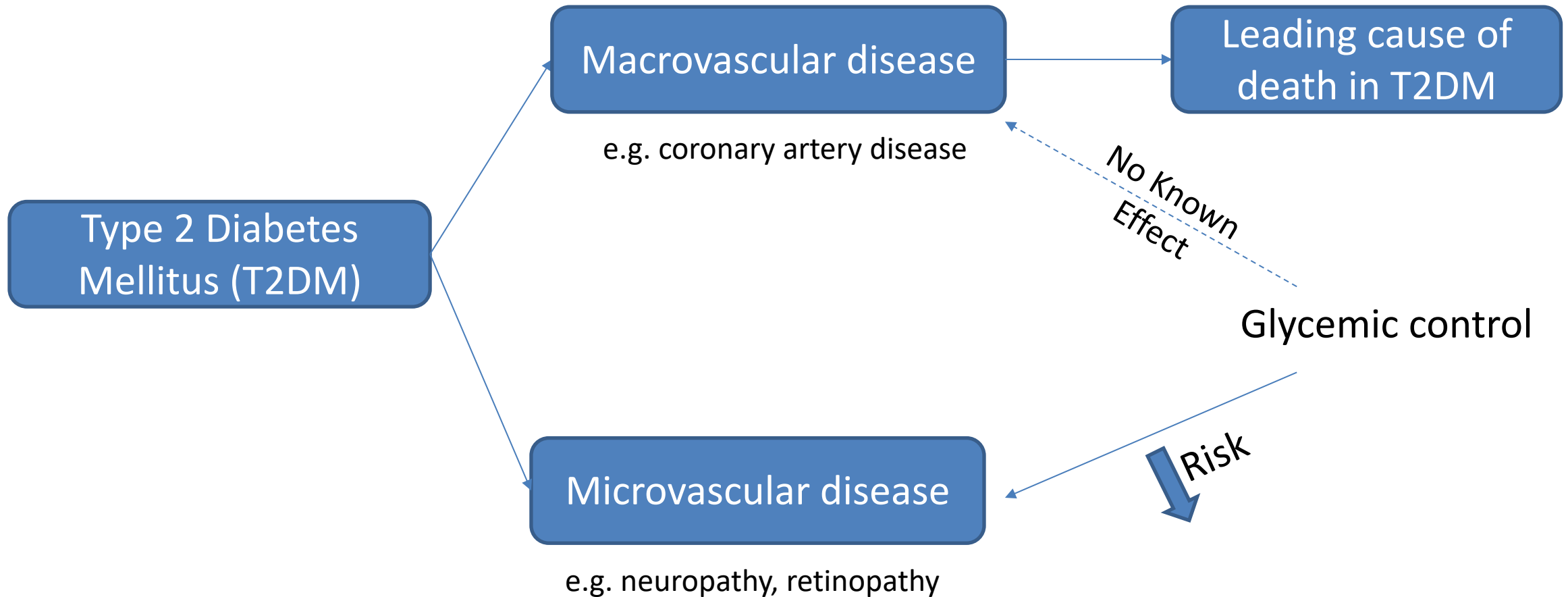
LEGEND-T2DM: Second-line antihyperglycemic treatment protocol

Rohan Khera

on behalf of the LEGEND initiative



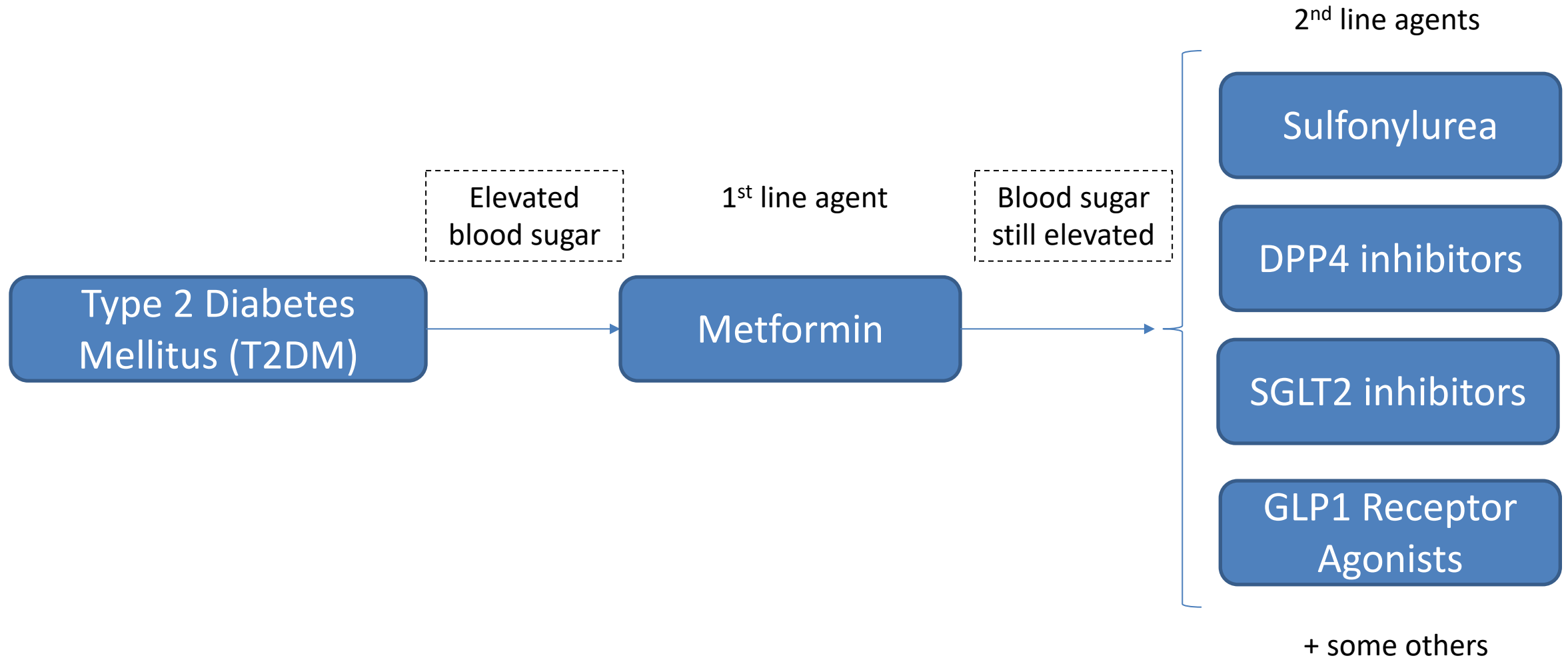
What is the Motivation for LEGEND-T2DM?





What is the Motivation for LEGEND-T2DM?

Several treatment strategies

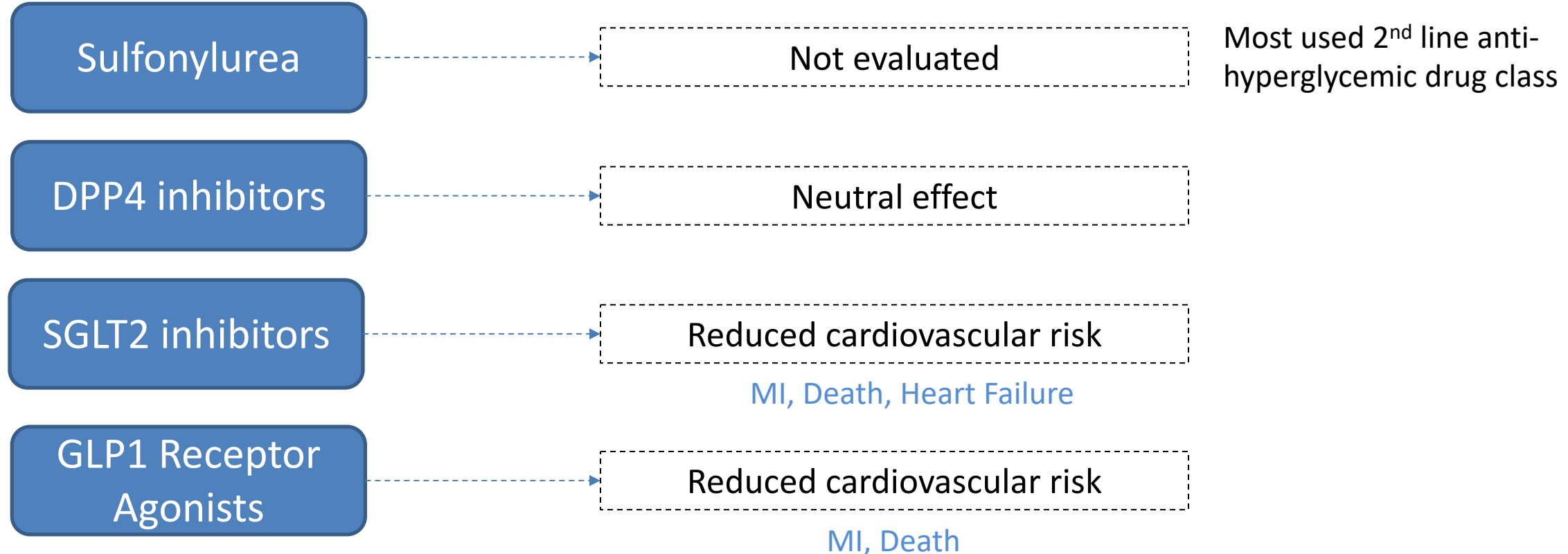




What is the Motivation for LEGEND-T2DM?

Variable evidence for cardiovascular efficacy for agents

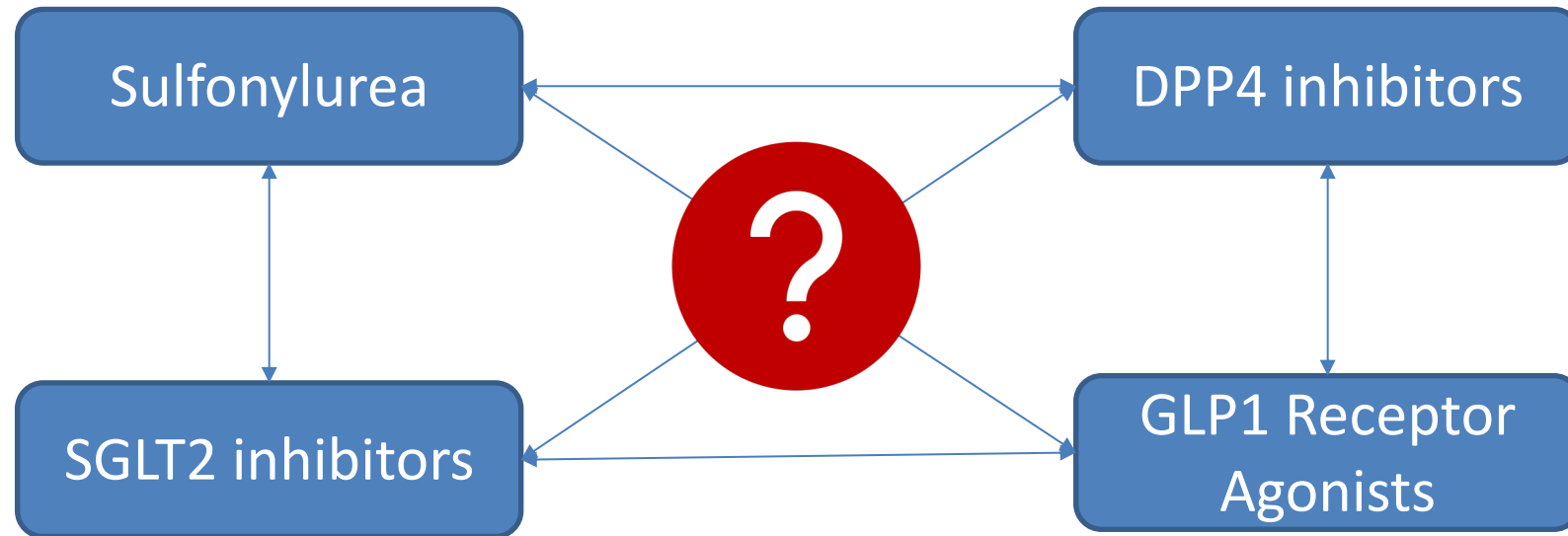
Cardiovascular outcome RCTs





What is the Motivation for LEGEND-T2DM?

Unknown cardiovascular effects of agents relative to each other



Critical need for evidence to improve choice between drug classes



What is the Motivation for LEGEND-T2DM?

Unknown cardiovascular effects of agents relative to each other

Goals of LEGEND-T2DM

**Evaluate relative cardiovascular effectiveness and safety of
2nd line anti-hyperglycemic drug classes**

**Assess relative cardiovascular effectiveness and safety of
individual 2nd line anti-hyperglycemic agents across classes**



LEGEND-T2DM: Emerging best practices

Aki Nishimura and Fan Bu
on behalf of the LEGEND initiative

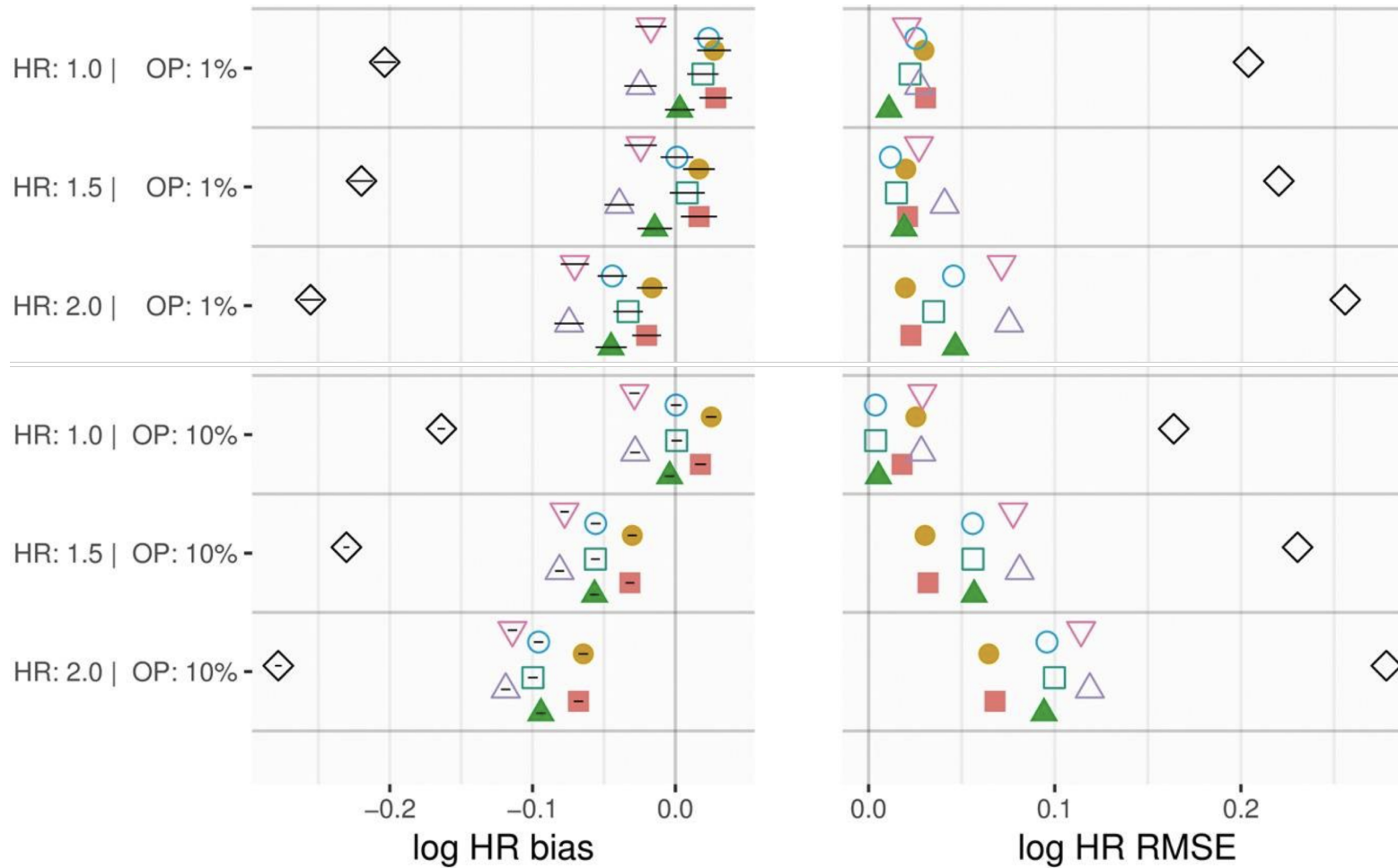


Methodological gaps in realizing LEGEND objectives

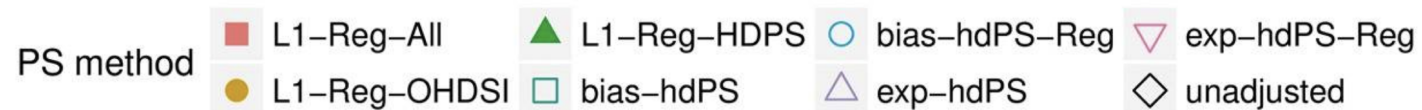
- Improved causal inference for survival outcomes
- Evidence synthesis of federated data sources



Causal inference on survival outcomes



Non-linearity /
non-collapsibility of
Cox model causes *bias*
toward null.





Causal inference on survival outcomes

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Volume 71, Issue 3
December 1984

Biased estimates of treatment effect in randomized experiments with nonlinear regressions and omitted covariates

M. H. GAIL, S. WIEAND, S. PIANTADOSI

Biometrika, Volume 71, Issue 3, December 1984, Pages 431–444, <https://doi-org.proxy1.library.jhu.edu/10.1093/biomet/71.3.431>

Published: 01 December 1984 **Article history**

 Springer Link

[Published: 18 January 2013](#)

On collapsibility and confounding bias in Cox and Aalen regression models

[Torben Martinussen](#)  & [Stijn Vansteelandt](#)

[Lifetime Data Analysis](#) **19**, 279–296 (2013) | [Cite this article](#)

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Causal inference on survival outcomes

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Volume 71, Issue 3
December 1984

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Optimal full matching for survival outcomes: a method that merits more widespread use

Peter C. Austin & Elizabeth A. Stuart,

First published: 06 August 2015 | <https://doi-org.proxy1.library.jhu.edu/10.1002/sim.6602> | Citations: 32

onfounding bias in Cox
models

[Torben Martinussen](#) & [Stijn Vansteelandt](#)

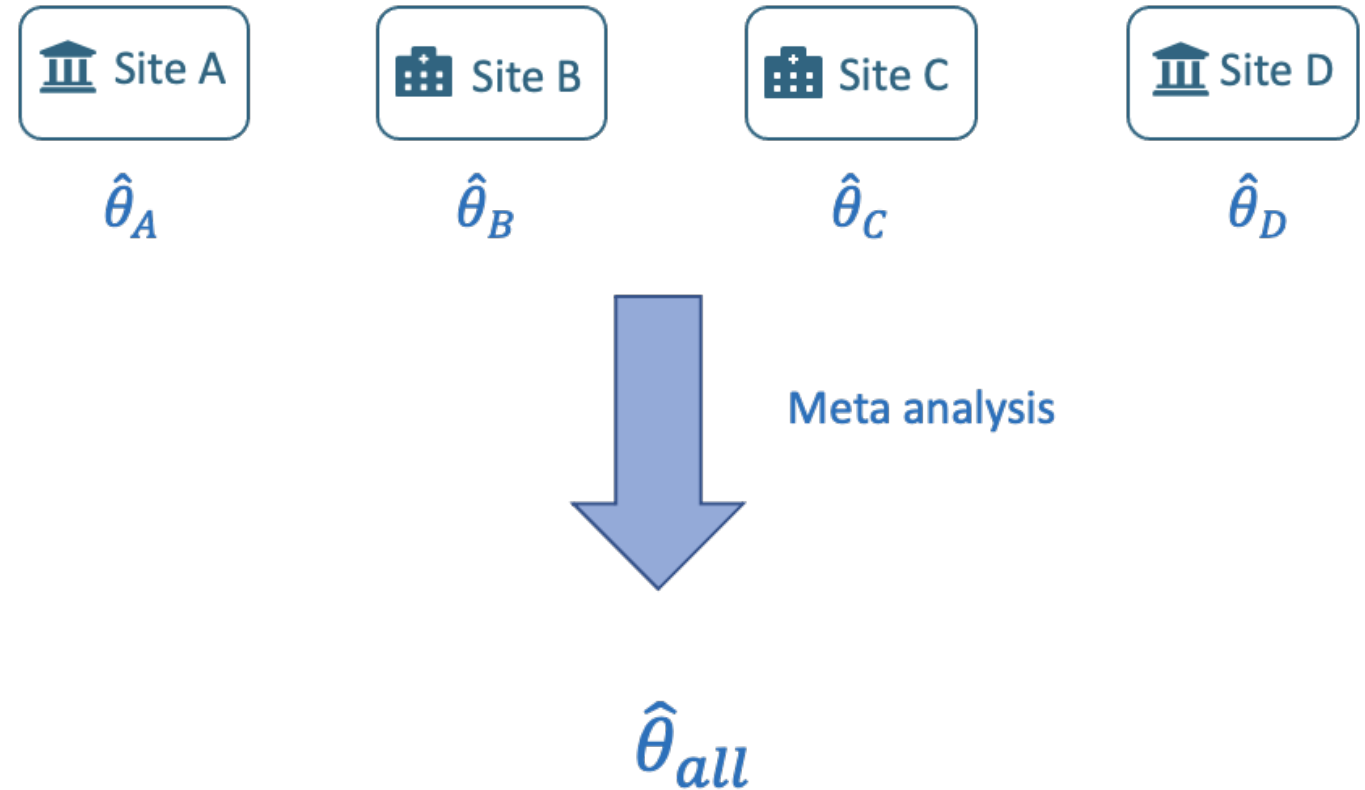
[Lifetime Data Analysis](#) **19**, 279–296 (2013) | [Cite this article](#)

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Evidence synthesis: traditional method

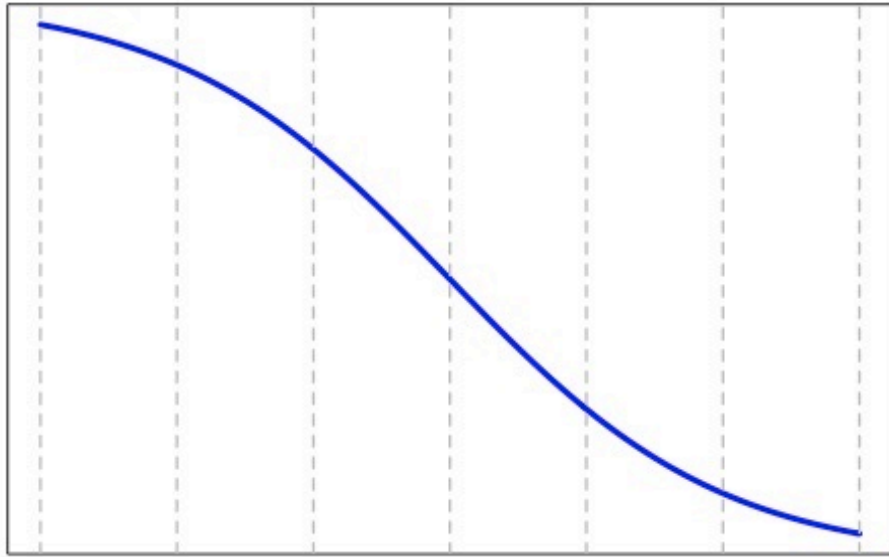
- Each site produces **point estimate** for effect size
- Use meta-analysis (e.g., random effect model) to combine estimates
- End result is an overall estimate





But there might be issues

Rare events with small or zero counts



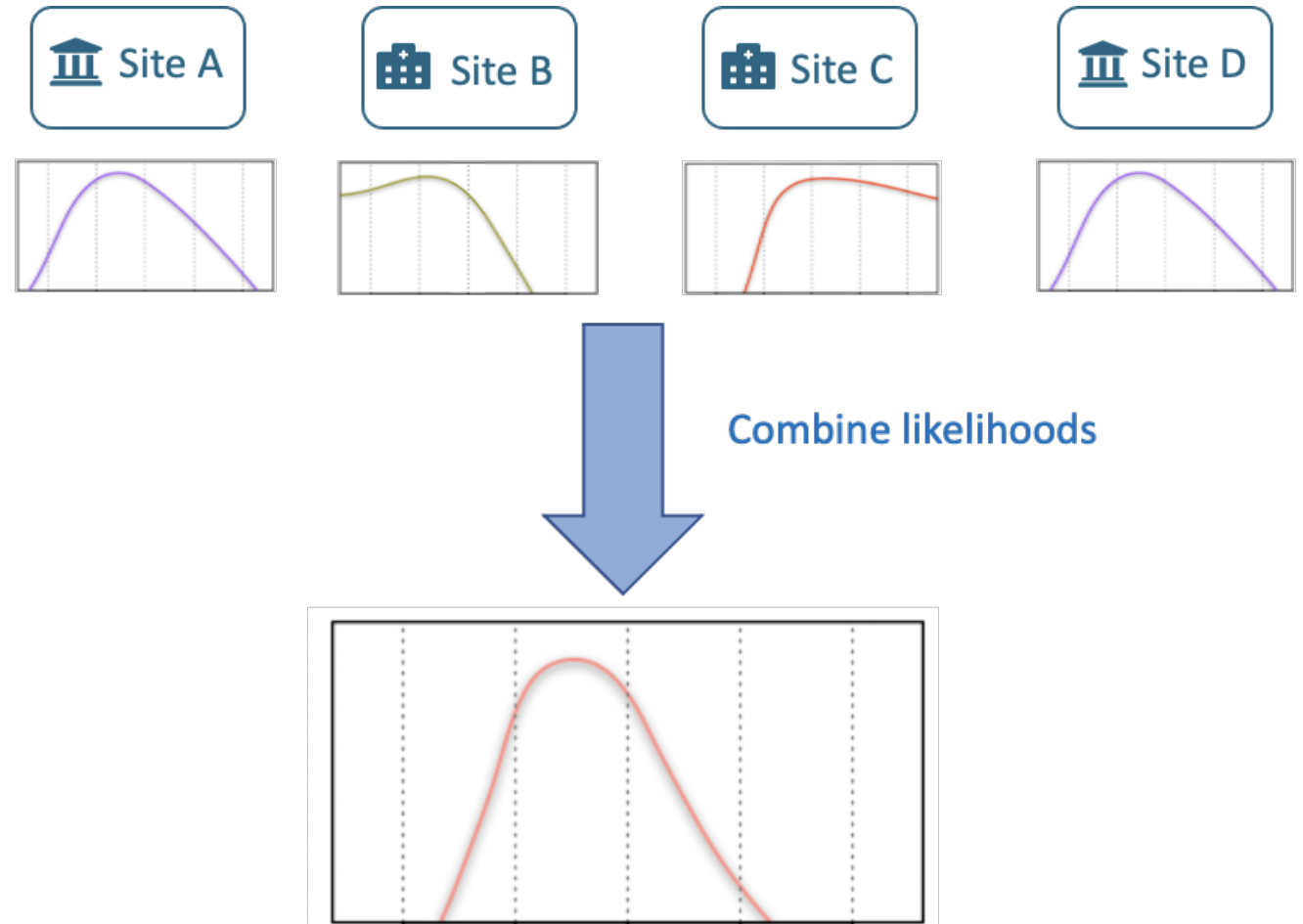
Likelihood monotonic:
no optimality!

- no MLE, no confidence interval
- traditional method **discards** such data sites



Evidence synthesis: emerging alternative

- Instead of summary statistics
- Each site produces a **likelihood**
- Likelihood: profile of full evidence
- Use Bayesian or numerical methods to "combine" likelihoods
- End result is an overall likelihood profile





Challenges and emerging & future work

- Small or zero counts (monotonic individual likelihoods)
 - need better likelihood approximations or better combination algorithms (e.g., Schuemie et al., 2021; Duan et al., 2020)
- Systematic errors (common in observational data)
 - calibration using negative (& positive) controls (e.g., Mulgrave et al., 2020)
- Community-level confounding
 - possible solution with mixture models or latent factor models



LEGEND-T2DM: How to get involved?

Marc Suchard

on behalf of the LEGEND initiative



LEGEND-T2DM Kick-off!

- ENCePP EU PAS Register #43551
- Preliminary exposure and outcome cohort diagnostics:
 - Large US claims data sources: **CCAE, MDCR, MDCD**
 - Large US EHR data sources: **VA, OptumEHR**
 - US academic EHR data sources: **Columbia, Hopkins**
 - Int'l EHR data sources: **NUS (Singapore)**
- Formal study start date: 1 Nov 2021 !!

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BMJ Yale

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RESEARCH PROTOCOL: Large-scale evidence generation and evaluation across a network of databases for type 2 diabetes mellitus

[Rohan Khera](#), [Martijn J Scheumie](#), [Yuan Lu](#), [Anna Ostroplets](#), [Ruijun Chen](#),
[George Hripacsak](#), [Patrick B Ryan](#), [Harlan M Krumholz](#), [Marc A Suchard](#)
doi: <https://doi.org/10.1101/2021.09.27.21264139>

This article is a preprint and has not been peer-reviewed [what does this mean?]. It reports new medical research that has yet to be evaluated and so should not be used to guide clinical practice.

Abstract

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Abstract

Background Therapeutic options for type 2 diabetes mellitus (T2DM) have expanded over the last decade with the emergence of sodium-glucose co-transporter-2 (SGLT2) inhibitors and glucagon-like peptide-1 (GLP1) receptor agonists, which reduced the risk of major cardiovascular events in randomized controlled trials (RCTs). Cardiovascular evidence for older second-line agents, such as sulfonylureas, and direct head-to-head comparisons, including with dipeptidyl peptidase 4 (DPP4) inhibitors, are lacking, leaving a critical gap in our understanding of the relative effects of T2DM agents on cardiovascular risk and on patient-centered safety outcomes.



LegendT2dm Study Package

ohdsi-studies / LegendT2dm Public

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Code Issues 1 Pull requests 1 Actions Projects Wiki Security Insights Settings

master 3 branches 0 tags

Go to file Add file Code

msuchard	Update start date	✓ ed876f8 · 3 minutes ago	205 commits
Documents	add EUPAS number	4 days ago	
R	clean error in file name	4 days ago	
docs	add EUPAS number	4 days ago	
extra	clean error in file name	4 days ago	
inst	manual merge from debug	5 days ago	
man	standardize result file names and parameterize more for in...	4 days ago	
.Rbuildignore	remove many code warnings and notes	2 months ago	
.gitignore	better gitignore	22 days ago	
DESCRIPTION	remove many code warnings and notes	2 months ago	
LegendT2dm.Rproj	first commit	15 months ago	
NAMESPACE	standardize result file names and parameterize more for in...	4 days ago	
README.md	Update start date	3 minutes ago	
renv.lock	update README and results uploading (untested)	11 days ago	

README.md

Large-scale Evidence Generation and Evaluation across a Network of Databases for Type 2 Diabetes Mellitus (LEGEND-T2DM)

Study Status Results Available

- Analytics use case(s): Population-Level Estimation
- Study type: Clinical Application
- Tags: -
- Study lead: Marc A. Suchard
- Study lead forums tag: msuchard
- Study start date: 1 November 2021
- Study end date: -
- Protocol: [HTML document](#)
- Publications: -
- Results explorer: -

Requirements

- A database in [Common Data Model version 5](#) in one of these platforms: SQL Server, Oracle, PostgreSQL, IBM Netezza, Apache Impala, Amazon RedShift, Google BigQuery, or Microsoft APS.
- R version 4.0.5
- On Windows: [RTools](#)
- [Java](#)
- 100 GB of free disk space

About

No description, website, or topics provided.

Readme

Releases

No releases published
[Create a new release](#)

Packages

No packages published
[Publish your first package](#)

Contributors 4

- msuchard Marc Suchard
- rohanxera Rohan Khara
- schuemie Martijn Schuemie
- aostroplets Anna Ostroplets

Environments 1

github-pages Active

Languages

R 69.2% · TeX 30.7% · CSS 0.1%

- Distributed as an OHDSI Network study R package from github.com
 - OMOP CDM v5 and R 4.0.5
 - README.md
 - renv / dockerfile
- Cohort characterization is ready for execution across our network
- Please contribute as a *data partner*



LEGENDary Resources

- Team:

- | | |
|-------------------|--------------------|
| – Fan Bu | – Evan Minty |
| – Ray Chen | – Aki Nishimura |
| – George Hripcsak | – Anna Ostropolets |
| – Rohan Khera | – Patrick Ryan |
| – Harlan Krumholz | – Martijn Schuemie |
| – Kelly Li | – Marc Suchard |
| – Yuan Lu | |

- Join us:

- 2nd/4th/5th Thursday of each month at 12p ET



- Links:

- Study protocol: <https://ohdsi-studies.github.io/LegendT2dm/Protocol.html>
- Study package: <https://github.com/ohdsi-studies/LegendT2dm>

- Contact to participate: **Marc Suchard** (Teams) or msuchard@ohdsi.org