





Leveraging Post-marketing Drug Safety Research through Semantic Technologies: The PharmacoVigilance Signal Detectors Ontology

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Background: How is Drug Safety Being Assessed?

- **Pre-market** setting: clinical trials of drugs
 - Limitations: Time constraints, population size and potential bias

not sufficient to detect all possible safety issues

• **Post-market** surveillance: continuous monitoring throughout the time a drug is actively prescribed

many heterogeneous data sources to investigate (e.g. spontaneous reports, observational healthcare data, literature, social media, etc.)

Definitions (in the scope of this work)

- **Signal** (in Pharmacovigilance):
 - Any information (reported or computationally extracted) on an unknown / incompletely documented possible causal relationship between a drug and an adverse effect

Signal Detection Method:

- Computational method aiming to identify pharmacovigilance signals
- Designed to operate on a specific type of data (i.e. contingency tables of drugcondition pairs)
- Offers a **set of parameters / analysis options**, e.g. thresholds for decision making, ranking criteria, etc.
- Results in a ranked list of drug-condition pairs along with features denoting their statistical significance

Signal Detector:

 A software implementation of a signal detection method executed based on a specific parameterization

Example: The GPS¹ Implementation (PhViD²) Input Data & Parameters

Usage

```
GPS(DATABASE, RR0 = 1, MIN.n11 = 1, DECISION = 1, DECISION.THRES = 0.05, RANKSTAT = 1, TRONC = FALSE, TRONC.THRES = 1, PRIOR.INIT = c(alpha1 = 0.2, beta1 = 0.06, alpha2 = 1.4, beta2 = 1.8, w = 0.1), PRIOR.PARAM = NULL)
```

Arguments

DATABASE Object returned by the function as . PhViD.

RR0 Value of the tested risk. By default, RR0=1.

MIN.n11 Minimum number of notifications for a couple to be potentially considered as a

signal. This option does not affect the calculation of the hyper parameters. By

default, MIN.n11 = 1.

• • •

¹Gamma Poisson Shrinker, used by FDA to screen reports gathered in FAERS. Details: W. DuMouchel, "Bayesian data mining in large frequency tables, with an application to the FDA spontaneous reporting system", The American Statistician, 53, 1999, 190–6.

²I. Ahmed and A. Poncet, "PhViD: an R package for PharmacoVigilance signal Detection", http://cran.r-project.org/web/packages/PhViD/

Example: GPS Implementation (PhViD) Output Options

Value

ALLSIGNALS Data.frame summarizing the results of all couples with at least MIN.n11 noti-

fications ordered by RANKSTAT. It contains notably the labels, the cell counts, the expected counts, RANKSTAT, the ratios(count/expected count), the marginal

counts and the estimations of FDR, FNR, Se et Sp. If RANKSTAT!=1, the last

column is the posterior probability of the null hypothesis.

SIGNALS Same Data.frame as ALLSIGNALS but restricted to the list of generated signals.

NB. SIGNALS Number of generated signals.

INPUT. PARAM Parameters entered in the function.

PARAM A list that contains the prior hyper parameters (PRIOR.PARAM). Additionally if

PRIOR.PARAM=NULL, it also contains the prior hyper parameters initialization

(PRIOR.INIT) and the convergence code (see nlm()).

Our Ultimate Goal (SAFER project)

 Develop a platform capable of exploiting and complementing evidence obtained from diverse, existing signal detection methods and data sources

• <u>Why</u>:

- Uncertainty in the results obtained from all types of computational signal detection methods
- Accurate indications & timely decisions required

Position statement:

- Each method and data source may contribute at a different level in complementing our knowledge on drug safety risks
- → An integrative perspective may add value!

Some Challenges Introduced by the Integrated Perspective

Practical Perspective:

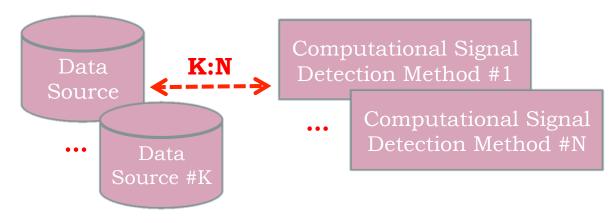
• Are **sufficient and appropriate methods and data** available to implement the integrated approach?

- <u>Technical Perspective</u>:

- How can this integration be systematic and large-scale?
- How to obtain a combined, prioritized list of the findings?

End-user Perspective:

- How can I use all these methods without having to be an expert in statistics?
- How can I setup an analysis experiment based on my interest?
- How can I make my results understandable and easy to share with other researchers?





Practical Perspective: Availability of <u>Data Sources</u> for Signal Detection



Trend toward the provision of programmatic access to data sources

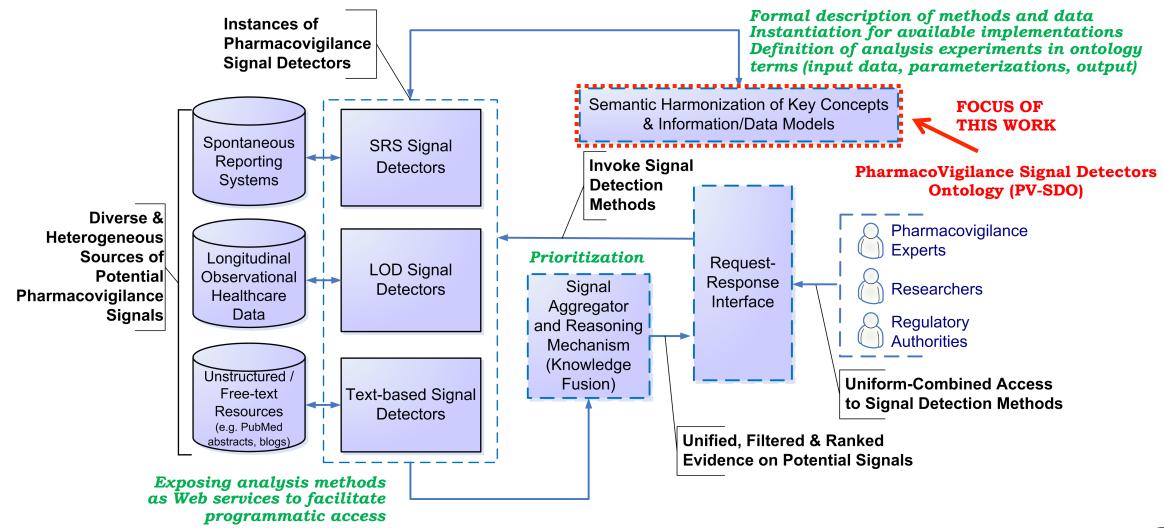


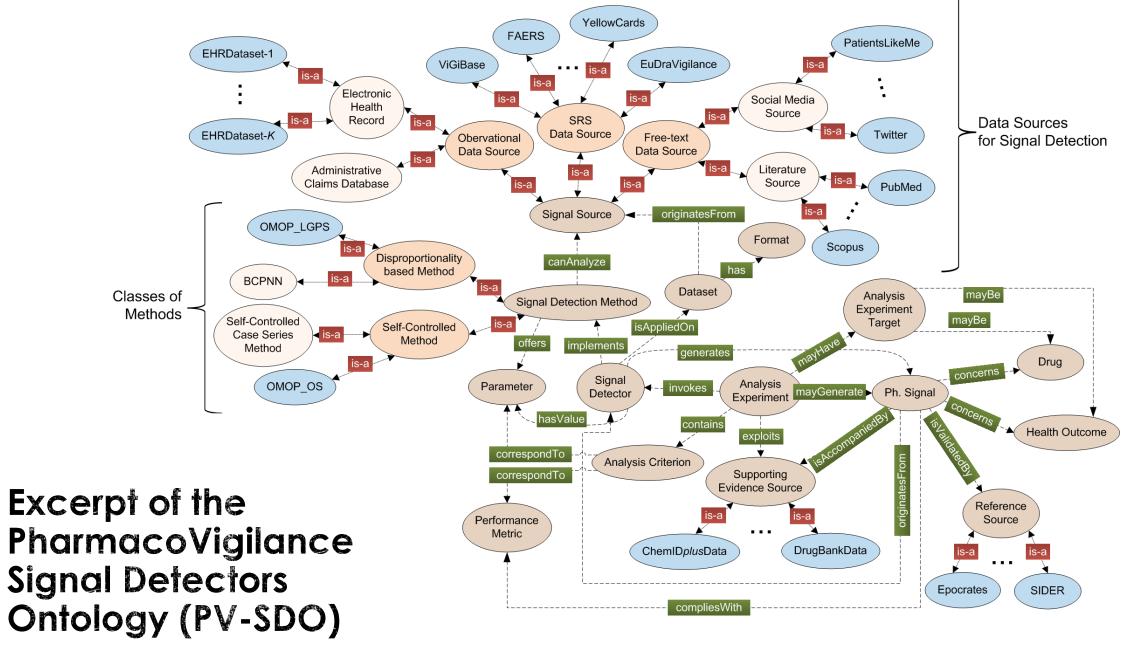
Practical Perspective: Availability of Open-Source Signal Detection <u>Methods</u> - *List of OMOP Implementations*¹

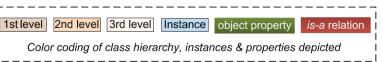
- Disproportionality Analysis (DP) OMOP Research Team
- Univariate Self-Controlled Case Series (USCCS) OMOP Research Team
- Observational Screening (OS) ProSanos Corporation
- Multi-Set Case Control Estimation (MSCCE) OMOP Research Team
- Bayesian Logistic Regression (BLR) OMOP Research Team
- Case Control Surveillance (CCS) Lilly
- IC Temporal Pattern Discovery (ICTRL) Uppeara Monitoring Centre
- Case-Crossover (CCO) Joy rsity of Stah
- HSIU Population-Based Method Indiana University
- Maximized Sequential Probability Ratio Test (MSPRT) Harvard Pilgrim
- Conditional Sequential Sampling Procedure (CSSP) Harvard Pilgrim
- High-Dimensional Propensity Score (HDPS) OMOP Research Team
- Incident User Design (IUD-HOI) M. Alan Brookhart

¹http://omop.org/MethodsLibrary

Technical Perspective: Systematic, Large-scale Integration and Prioritization of Outcomes (the SAFER framework)







End-user Perspective: Hiding Complexity

Semantic harmonization of domain concepts:

- classification and annotation of data sources for signal detection (either for discovery or filtering)
- classification of signal detection methods
- classification of analysis parameters, performance metrics and ranking criteria
 that can be set/used by signal detection methods
- Facilitate the **definition of drugs and health outcomes of interest** through semantic mappings

Semantic Harmonization Classification of Signal Detection Method Implementations (excerpt)

- ▼ ⑤ 'Signal Detection Method'
 - ▼ 'Bayesian Confidence Propagation Neural Network'
 - 'Bayesian Confidence Propagation Neural Network Extended To The Multiple Comparison Framework'
 - 'Original Bayesian Confidence Propagation Neural Network Original'
 - 'Case Control Signal Detection Method'
 - 'Cohort Signal Detection Method'
 - Gamma Poisson Shrinkage based Signal Detection Method
 - 'Gamma Poisson Shrinkage Extended To The Multiple Comparison Framework'
 - 'Longitudinal Gamma Poisson Shrinker'
 - 'Multi-item Gamma Poisson Shrinker'
 - 'Original Gamma Poisson Shrinkage'
 - Proportional Reporting Ratio based Method
 - 'Original Proportional Reporting Ratio Signal Detection Method'
 - 'Proportional Reporting Ratio Method Extended To The Multiple Comparison Framework'
 - 'Reporting Fisher's Exact Test Signal Detection Method'
 - Reporting Odds Ratio based Method'
 - 'Original Reporting Odds Ratio Method'
 - 'Reporting Odds Ratio Method Extended To The Multiple Comparison Framework'
 - Self-Controlled Signal Detection Method
 - 'Self-Controlled Case Series Signal Detection Method'
 - 'Self-Controlled Cohort Design Signal Detection Method'

Indentation denotes class-subclass relation



Semantic Harmonization Classification of Analysis Parameters (excerpt)

- Parameter
 - 'Apply Leopard Filtering For Protopathic Bias'
 - 'Case-Control Matching Strategy'
 - 'Comparator Population'
 - Control Related Parameter
 - 'Control Period'
 - 'Controls PerCase'
 - 'Include Index Date In Control Period'
 - 'Use Control Period In Expected Calculation'
 - Covariate Related Parameter
 - 'Additional Covariates Included In The Propensity Score Model'
 - 'Covariate Eligibility Window'
 - 'Covariate Selection Algorithm Additional Parameters'
 - 'Dimensions To Include As Potential Covariates'
 - Decision Related Parameter'
 - 'Decision Rule For The Signal Generation'
 - 'Threshold for Decision'
 - 'Exposures To Include'
 - 🧶 'Include All Drugs In A Multivariate Analysis'
 - 'Include Index Date In Time At Risk'
 - 'Minimum Number Of Notifications For A Couple To Be Potentially Cor
 - 'Nesting Within Population With The Indication Of The Target Drug'
 - 'Outcomes To Include'
 - Prior Related Parameter
 - ▼ 'Period Prior To Exposure In Expected Calculation'
 - 'Use One Day Prior To Exposure In Expected Calculation'
 - 'Use One Month Prior To Exposure In Expected Calculation'
 - 'Prior Distribution'
 - 'Prior Parameters Chosen'
 - 'Prior Parameters Initialization Vector'
 - 'Required Observation Time Prior To Exposure'
 - 'Required Observation Time Prior To Outcome'
 - 'Variance Of The Prior'

End-user Perspective: Facilitate Study Setup Example: Mapping Terms across Diverse Data Sources

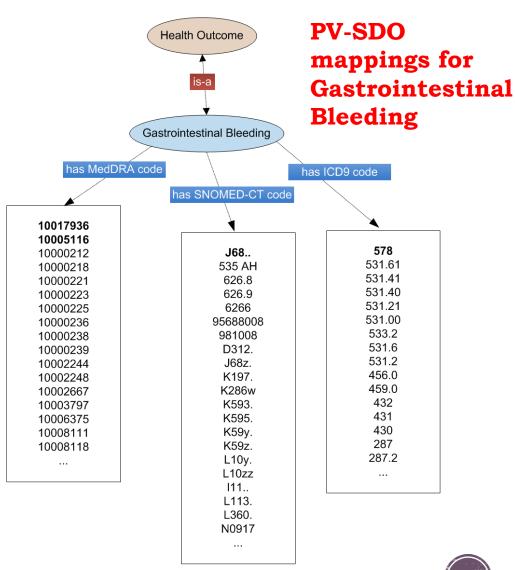
Illustration of the problem

Table 1 AERS and EHR—event and outcome definitions

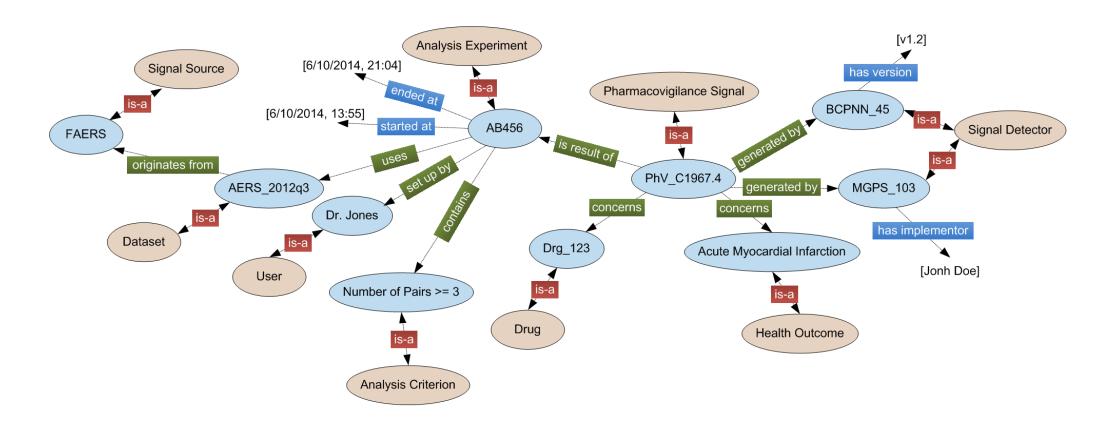
Event	AERS MedDRA PT event definitions	EHR Outcome definitions
Pancreatitis	Pancreatitis acute	Lab tests: amylase >300 U/I or lipase >120 U/I
Rhabdomyolysis	Rhabdomyolysis, blood creatine phosphokinase MM increased	Lab test: 5× normal levels of creatine kinase
ατ	Long QT syndrome, ECG QT prolonged, Torsade de pointes, ECG QT interval abnormal, ventricular tachycardia	UMLS codes: C0023976, C0151878, C0743431, C0855333, C1560305

AERS, adverse event reporting system; EHR, electronic health record; PT, preferred term.

<u>Table source:</u> R. Harpaz et al., Combing signals from spontaneous reports and electronic health records for detection of adverse drug reactions, J Am Med Inform Assoc. 2013;20(3):413-9.



End-user Perspective: PV-SDO Annotation of Experiments – Results Sharing



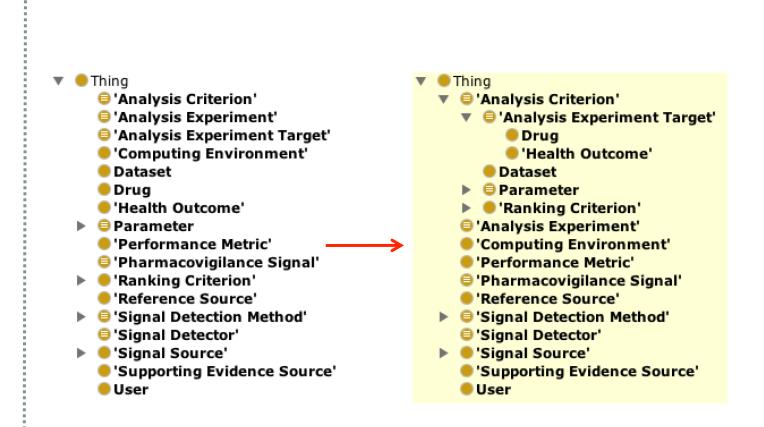
Example: Provenance of analysis outcomes (fictive data)



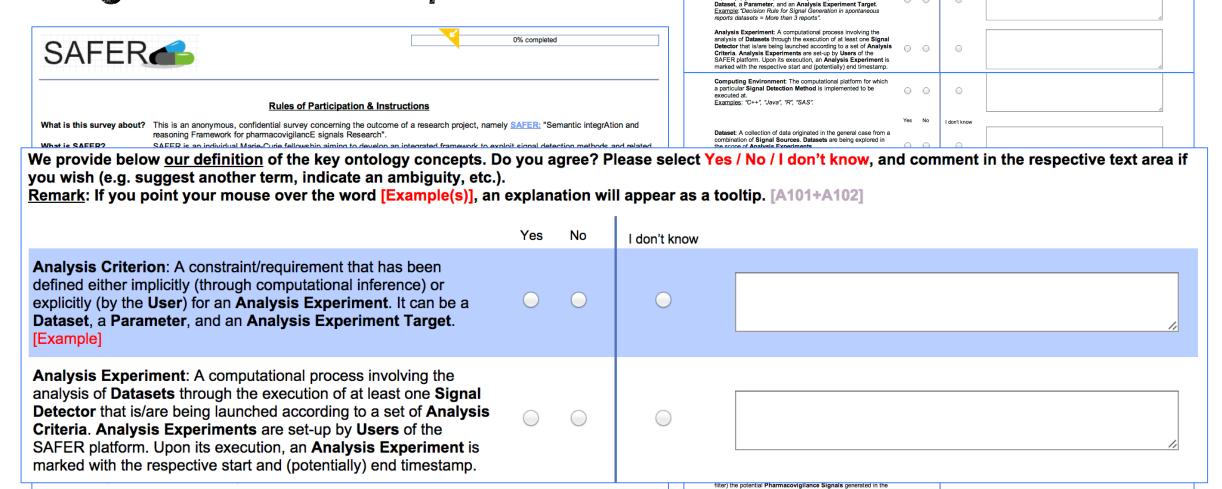
Results: PV-SDO Evaluation Data-driven & Automatically Using a Reasoner

Is the ontology sufficient to describe methods which were not part of the source knowledge employed in the design?

Test with new published methods



Results: PV-SDO Evaluation By Knowledge Engineers & Signal Detection Experts



Online survey

Part I: Questions on Main Concept Definitions in the Signal Detectors Ontology

We provide below our definition of the key ontology concepts. Do you agree? If not, please comment on the respective text area. [A101+A102]

0 0

0 0

Analysis Criterion: A constraint/requirement that has been defined either implicitly (through computational inference) or explicitly (by the User) for an Analysis Experiment. It can be a

scope of an Analysis Experiment. This class is specialized into

two types, Domain-specific Ranking Criterion and

Current version of PV-SDO: 1,312 axioms, 101 classes, 34 object properties, 32 data properties and 168 individuals

Discussion: Foreseen Value of Combining Data Sources & Detection Methods

- Support the detection of true associations reduce false positive findings through replicated signaling;
- Support timely decisions: the "many sources many methods" setting may provide faster indications;
- Address the event-based and data-source based differential performance of methods;
- Address the fact that no individual approach to detect signals is adequate and the concurrent use of multiple methods is essential,

by hiding the complexity for end-users (i.e. drug safety experts and regulatory authorities)

Next Steps

PV-SDO:

- Complete revisions based on signal detection experts' feedback
- Investigate **extension** with domain knowledge, e.g. for method selection
- Provide a comprehensive resource for end-users

PV-SDO availability:

<u>http://bioportal.bioontology.org/ontologies/PV-SDO</u>, currently private access – publicly available soon!

- Focus on favorable **test cases** (adverse effects and/or drugs) and **scenarios** (e.g. targeted vs. strong signal identification)
- A proof-of-concept end-to-end integration and evaluation





Discussion

<u>Acknowledgement</u>

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Thank you!

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