# Subset disproportionality analysis within a global database to uncover adverse drug reactions in risk groups

Lovisa Sandberg, Yasunori Aoki, Rebecca Chandler, Henric Taavola and G. Niklas Norén, Uppsala Monitoring Centre, Uppsala, Sweden



#### Example of a subset disproportionality analysis

1

Information Component (IC) measures of disproportionality for a specific drug and adverse event, across different data subsets, with 95% credibility intervals for the overall analysis and 99% credibility intervals for the subsets to avoid highlighting spurious associations. [1]

Α	All	Age	Sex	BMI	Pregnancy	Underlying condition	Continent	Country
4 -			Ŧ	Ţ		T T		т т Т
IC 0-	<u>₹</u>		<u>+</u>					
-4		<ul> <li>≤ 27 days to</li> <li>28 days to</li> <li>23 months</li> <li>2 to 11 years</li> <li>1 to 17 years</li> <li>1 to 44 years</li> <li>45 to 64 years</li> <li>65 to 74 years</li> <li>5 75 years</li> </ul>	Female - Male -	Obese adult - Underweight _ adult _	Pregnant -	Coronary artery disorders Embolism and thrombosis skin appendage conditions conditions disorders ()	Africa - Asia - Europe - North America - ()	Australia - Canada - Sweden - South Africa - ()

# Background



There is growing recognition of the variability between patients in both the benefits and harms from medicinal products. In pharmacovigilance, current statistical methods to screen large data sets are sensitive to associations at the population level but can fail to highlight suspected adverse drug reactions (ADRs) in sub-populations at risk.

## Methods

Dataset: 15.4 million reports retrieved on 28 August 2017 from VigiBase, the WHO global database of individual case safety reports Disproportionality analyses performed for drug-adverse event (AE) Subset pairs (1) in the entire database and (2) across a range of data subsets. disproportionality Drug-AE pairs disproportionally overreported in such subsets but not in the whole data, and with the observed-to-expected ratio in the subset analysis at least twice that in the whole data, were identified. Identified drug-AE-subset associations ordered by (1) **Prioritization** vigiRank [2] for strength of evidence, and (2) weighted random sampling for subset balancing. SS Manual review of top-ordered drug-AE-subset associations including review of the reports and Initial review consultation of literature in search for support for

To explore if disproportionality analysis across subsets of individual case reports within a global database can uncover signals of suspected adverse drug reactions in certain risk groups.

# Results

#### Initial review

Out of 386 manually reviewed drug-AE-subset associations, **18** (4.6%) were classified as **potential signals**. The highest yield was identified in **females** (5), **underweight** adults (3), and the **elderly** (3).

As of August 2018, in-depth clinical

reviews have been completed for 14

out of 18 potential signals, resulting

in **seven signals** describing

potential risk groups for ADRs.

Covariate	Potential signals	Assessed associations	Proportion
Sex	6	44	14%
BMI	4	39	10%
Continent	4	77	5%
Age	3	65	5%
Country	1	62	2%
Pregnancy	0	33	0%
Underlying condition	0	75	0%

**1** 

#### Signals of ADRs in risk groups

Aflibercept – Deep vein thrombosis and pulmonary embolism – **Males** 

Ceftriaxone – Hepatitis – 75 years and older
 Esomeprazole – Gynaecomastia – Obese
 Glibenclamide – Palpitations – Asian population
 Levofloxacin – Myoclonus – 75 years and older
 Omalizumab – Anaphylactic shock – Females
 Selegiline – Hypoglycaemia – Underweight

#### In-depth review

Signals of ADRs in risk groups

In-depth review of potential signals performed by clinical experts.

possible risk group.

### Conclusions

d

σ

Σ

In-depth review

Signals of suspected adverse drug reactions in risk groups can be identified through subset disproportionality analysis within a global database. Further development of such methods could usher in a new era of "precision pharmacovigilance".

#### References

[1] Hopstadius J, Norén GN. Robust discovery of local patterns: Subsets and stratification in adverse drug reaction surveillance. Proceedings of the 2nd ACM SIGHIT International Health Informatics Symposium ACM: Miami, FL, 2012: 265–274.

[2] Caster O, Juhlin K, Watson S, Norén GN. Improved statistical signal detection in pharmacovigilance by combining multiple strength-of-evidence aspects in vigiRank. Drug Saf. 2014;37(8):617-28.

#### Disclosure

The authors are indebted to the national centres that contribute data to the WHO Programme for International Drug Monitoring. However, the opinions and conclusions in this study are not necessarily those of the various centres, nor of WHO.

Uppsala Monitoring Centre

**Uppsala Monitoring Centre (UMC)** Box 1051, SE-751 40 Uppsala, Sweden

+46 18 65 60 60 www.who-umc.org