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Background										
•	<b>Upper gastrointestinal bleeding (UGIB)</b> is a serious medical emergency leading to death in about 10% of cases.									
•	The French Nationwide Healthcare System database (SNDS) covers the overall French population from birth to death (66.6 million people). It includes individual pseudonymised information on all reimbursed healthcare expenditures, including drugs, and hospital discharges summaries.									
		Objectives								
	Drug safety alert generation associated with UGIB may be achieved through the application of empirically validated and calibrated case-based methods in the SNDS.									
	The present work aims to identify the optimum design and settings for the identification of drugs associated with UGIB in the SNDS.									
		Methods								
•	156 057 UGIB cases were	extracted from SNDS over	<sup>-</sup> 2009-2014.							
•	<ul> <li>Positive controls: drugs with a known association with UGIB</li> <li>Negative controls: drugs with no known association with UGIB</li> <li>Controls with a minimal detectable relative risk ≤1.30 in the relevant population were deemed detectable and kept.</li> <li>96 SCCS, 20 CC and 80 CP variants were used to measure association between drug controls and UGIB in a 1/10<sup>th</sup> sample of the population (Table 1).</li> </ul>									
7	Table 1. Description of design value	riants Approaches								
	Self-controlled case series	Case-Control	Case-population							
Settings	Outcomes to include: <i>All occurrences / First occurrence</i> <b>Risk window:</b> <i>30 days following dispensing / Overall</i> <i>period covered by dispensing</i> <b>Pre-exposure window:</b> <i>0 day / 7 days / 30 days</i> <b>Age included into the model:</b>	Outcomes to include: <i>All occurrences / First occurrence</i> <b>Risk window:</b> <i>7 days / 30 days / 60 days</i> <b>Lag periods:</b> <i>0 day / 7 days / 15 days</i> <b>Controls matched per cases (on</b> age and gender):	Outcomes to include: All occurrences / First occurrence Risk window: 7 days / 30 days / 60 days Lag periods: 0 day / 7 days / 15 days Approach							
	Yes / No Seasonality included into the model: Yes / No All dispensed drugs included into the model (multiple drug use): Yes /No	Up to 2 / Up to 10	Count data (per-user) / person-time <b>Extrapolation of the aggregated</b> data: Raw (no stratification) / Stratified on age and gender <b>Measure of association</b> Case-population Ratio / predicter Relative Risk							
•	Yes / No Seasonality included into the model: Yes / No All dispensed drugs included into the model (multiple drug use): Yes /No Performance of each design the receiving operator cu	gn variants was assessed rve (AUC), the mean squa	Count data (per-user) / person-time Extrapolation of the aggregated data: Raw (no stratification) / Stratified on age and gender Measure of association Case-population Ratio / predicter Relative Risk based on the area under are error (MSE).							
•	Yes / No Seasonality included into the model: Yes / No All dispensed drugs included into the model (multiple drug use): Yes /No Performance of each design the receiving operator cu Parameters that had ma approach were identified - Dependent variable = p of the AUC distributions - Independent covariates	gn variants was assessed rve (AUC), the mean squa ajor impact on results through logistic regress probability that a variant had of the variants. = parameters that were variant	Count data (per-user) / person-time Extrapolation of the aggregated data: Raw (no stratification) / Stratified on age and gender Measure of association Case-population Ratio / predicter Relative Risk based on the area under are error (MSE). of the best performing ion: d an AUC >70th percentile aried in the variant.							

An empirical null distribution was derived from negative control estimates based on how often p < 0.05 while the null hypothesis was true, and used to calibrate *p*-value to take into account systematic and random error.





# **Empirical assessment of case-based methods for identification of drugs** associated with upper gastrointestinal bleeding in the French National Healthcare System database (SNDS)

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### Figure 2 shows that:

- 10 negative controls were significantly associated with UGIB; 4 positive controls were not significantly associated with UGIB.
- distribution Derived empirical null (supposed gaussian) had the following parameters:  $\mu$  =0.12;  $\sigma$  =0.17.
- Calibrating p-values (**Figure 3**)
- 2 negative controls were still significant: sucralfate and scopolamine;
- positive controls moved from significant to non-significant: potassium chloride, prednisolone, indomethacin, ibuprofen, fenoprofen, nabumetone, fluoxetine, citalopram, sertraline.

- SCCS considering the first outcome occurrence, adjusting for multiple drugs and using a 30-day risk window showed the best performances for drug-related UGIB assessment in the SNDS.
- Low systematic error seems to affect SCCS but protopathic bias and confounding by indication remained unaddressed issues.





Table 2. Univariate logistic regression analysis of self-controlled case series parameters influencing on the area under the receiver operating characteristics curve (AUC) in the 1/10<sup>th</sup> sampled population



## Conclusions

- increased the number of false negatives.

	Variants with low AUC n=59		Variants with high AUC n=37		High <i>vs.</i> Low AUC OR [IC à 95%]		p	AUC of the univariate model
							0.8375	0.51
ed	30	(50.8)	18	(48.6)	1			
	29	(49.2)	19	(51.4)	1.09	[0.48 - 2.48]		
,							0.8375	0.51
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	29	(49.2)	19	(51.4)	1.09	[0.48 - 2.48]		
				( )			0 0007	0.04
	20	$(\mathbf{C}1,\mathbf{O})$	10	(00.4)	4		0.0087	0.64
ences	36	(61.0)	12	(32.4)	1			
rence	23	(39.0)	25	(67.6)	3.17	[1.34 - 7.50]		
ıg use							<0.0001	0.80
ed	43	(72.9)	5	(13.5)	1			
	16	(27.1)	32	(86.5)	15.58	[5.30 - 45.77]		
re							0 4 4 0 4	0.00
							0.1404	0.62
	16	(27.1)	16	(43.2)	1			
	19	(32.2)	13	(35.1)	0.69	[0.26 - 1.86]		
	24	(40.7)	8	(21.6)	0.35	[0.12 - 0.99]		
N							<0.0001	0.73
	40	(67.8)	8	(21.6)	1			
om irst day	19	(32.2)	29	(78.4)	7.21	[2.80 - 18.54]		

AUC = area under the receiver operating characteristics curve; A high AUC was defined as an AUC≥0.75

Calibration process reduced the number of false positives but

ALCAPONE showed that SCCS with optimum settings has the potential to generate accurate UGIB-related drug safety alerts from SNDS, including hypotheses on its possible population impact.