Example 2: Adverse drug events detection and prevention by artificial intelligence



Data collection:

Denain General Hospital (Fr) Region Hovedstaden Hospitals (Dk)

Research Project:

PSIP european project Funded by the European Research Council Coordinator: Pr Régis Beuscart

Data analysis & software design:

E Chazard



PSIP

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Adverse drug events



Adverse drug events

- ADEs = Adverse Drug Events
- Several definitions. Institute of Medicine (2007):
 - "An injury resulting from the use of a drug"
 - "An injury due to medication management rather than the underlying condition of the patient"
- Epidemiological data:
 - 98,000 deaths per year in the US
 - An ADE would occur in 5-9% of inpatient stays
- Two fields of research:
 - Prospective ADE prevention
 - Retrospective ADE detection



Prospective prevention of ADEs



Adverse drug events prevention by CDSS based on level 1 artificial intelligence



Definitions

CPOE:

- computerized physician order entry
- process of electronic entry of medical practitioner instructions for the treatment of (hospitalized) patients

CDSS:

- Clinical decision support system
- Health information technology system that is designed to provide physicians and other health professionals with clinical decision support
- Often based on level 1 artificial intelligence (rules)
- CPOE + CDSS = the "obvious" solution for adverse drug events prevention?



CDSS & CPOE: Over-alerting, alerte-fatigue => poor clinical efficiency!

- **Over-alerting**: too numerous and inappropriate alerts
- Alerts interrupt the clinicians' workflow and induce alert-fatigue
 - Too many alerts
 - => time and mental energy consumption
 - => a mental state whereby users start ignoring critical alerts along with those that may be clinically insignificant
- May prevent CDSS from improving patient safety
- Alert override:
 - up to 96% of alerts are overridden by prescribers
 - But alert override is often inappropriate, and is sometimes followed by actual ADEs



What solutions have been implemented for over-alerting reduction?

- Support the medical management of the alert
 - Possible, but disappointing evaluation results [Dukes 2011, Dukes 2013]
- Expert filtering or tiering of the alerts, based on relevance
 - possible & efficient [Shah 2005, Van der Sijs 2008, Paterno 2009, Phansalkar 2013]
- Automated filtering or tiering of the alerts
 - Based on override statistics [Lee 2010]: less override but... is override decision reliable? [Slight 2013, Nanji 2014]
 - Based on outcome probability [Chazard 2009, Koutkias 2010, Chazard 2012]... proposed here



Criteria for assessing high-priority DDIs for clinical decision support in EHRs Phansalkar S - BMC med inf & decision making 2013

- 27 important criteria for choosing interactions to include for CDS in an EHR:
- Severity of interaction: Clinical Importance, likelihood of Mortality, of Morbidity, of Intervention
- Probability of interaction: Likelihood of the Adverse Reaction, timing of Administration, pharmacokinetic properties, Dose and Duration, Route, Sequence, Monitoring, Therapeutic window, Combination of drugs
- Clinical implications: Management burden, Monitoring planned for the interaction, Awareness of the intervention
- Patient characteristics: Alcohol, diet, smoking, drug use, Age, Gender, Concurrent diseases, Other active medications
- Evidence supporting interaction: Quantity of evidence, Quality of evidence, Biological plausibility

Similar conclusions in reviews or studies for overalerting reduction:

[Van der Sijs 2006] [Kuperman 2007] [Smithburger 2011] [Riedman 2011] [Ammenwerth 2011] [ung 2013]

Statistics-based contextualization of alerts (3)

Data-mining based segmentation of alert rules (8)

Retrospective ADE detection: a prerequisite for ADE prevention!

Idea driven by Pr Regis Beuscart, head of the PSIP Project



Funded by the European Research Council, 7th framework program (agreement N°216130)



Which methods for retrospective ADE detection ?

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Camp?	

- Reporting systems:
 - Based on spontaneous case reports
 - Mandatory, but underreporting bias: less than 5% cases are declared!
- Expert-operated chart reviews
 - Reference method, expert validation
 - Time consuming: 30 min per case, and some ADEs are very rare...





- Objective: using data reuse & data mining to:
 - Automatically identify past ADE cases
 - Generate ADE detection rules
 - Computing probabilities of occurrence

Retrospective detection of ADEs



Retrospective identification of past ADEs, although no explicit signal exists in the

Administrative data

88 years old woman

Diagnoses

- I10 Arterial hypertension
- Z8671 Personal history of myocardial ischemia
- I620 Non-traumatic subdural hemorrhage

Medical procedures

ABJA002 Drainage of an acute subdural hemorrhage, by craniotomy FELF001 Transfusion

Free-text reports

Surgical report

Discharge

letter

Acetaminophen

Drugs



Available data: ~175,000 inpatient stays from 6 hospitals (F, Dk, Bu)



Artificial intelligence Example of decision tree



 VKA= vitamin K antagonists (anticoagulant)

 INR= international normalized ratio.
 Evaluates VKA activity

- INR>5 => risk of hemorrhage
 - The tree attempts to explain INR>5

Artificial intelligence Example of decision tree

Rule #1 (4th leaf): **VKA** yes no f=0.1 **VKA** & butyrophenone Butyrophenone discontinuation discontinuation \rightarrow P=0.4 yes no f=0.4



Artificial intelligence Example of decision tree

Rule #2 (3rd leaf):



VKA & no butyrophenone discontinuation & hypoalbuminemia \rightarrow P=0.5



Validation of retrospective ADE detection

- Validation of each rule, on a bibliographic point of view
- II. Validation of the tool, on a statistical point of view



Artificial intelligence Expert validation of rules

Butyrophenone: neuroleptic drugs, may accelerate the intestinal transit



Artificial intelligence Expert validation of rules

Albumine = plasmatic protein to which VKA bind. Only the non-bound part is biologically active.

Serum albumin



Normal state: 99% of the VKA bind to albumin. Only 1% of VKA are biologically active. The intake is based on it.



O VKA

<u>Hypoalbuminemia:</u> decrease of the bound fraction, increase of the non-bound fraction => too high INR (with constant intake)

=> Need for validation, explanations, reorganization!



Overview of ADE detection rules

Kind of Outcome	# Rules
Coagulation disorders	
Hemorrhage (detected by the administration of haemostatic)	7
Heparin overdose (activated partial thromboplastin time>1.23)	5
VKA overdose (INR>4.9 or administration of vitamin K)	59
Thrombopenia (count<75,000)	24
Other coagulation disorders	23
Ionic and renal disorders	
Hyperkalemia (K ⁺ >5.3 mmol/l)	63
Renal failure (creatinine>135 µmol/l or urea>8 mmol/l)	8
Other ionic disorders	4
Miscellaneous	
Anemia (Hb<10g/dl)	2
Bacterial infection (detected by the administration of antibiotic)	4
Diarrhea (detected by the administration of an anti-diarrheal)	2
Fungal infection (detected by the administration of an antifungal)	10
Hepatic cholestasis (alk. Phos.>240 UI/l or bilirubins>22 µmol/l)	3
Hepatic cytolysis (ala. trans.>110 UI/l or asp. trans.>110 UI/l)	4
Hypereosinophilia (eosinophilocytes>10 ⁹ /l)	4
High level of pancreatic enzymes (amylase>90 UI/l or lipase>90 UI/l)	7
Neutropenia (count<1,500/mm ³)	2
Others	5
Total	236

2020-04-17

Pr Emmanuel Chazard - Artificial intelligence, Data reuse, Big data in Healthcare

Evaluation of the ADE retrospective detection

Complete 2010 year of one hospital

- Number of stays :
- Number of hyperkalemia cases :

14,747

117 (7.93‰) \rightarrow exhaustive review





Ability of the rules to retrospectively detect actual ADEs

Expert-operated case review

	Validated cases / reviewed cases	Precision [95% confidence interval]
Hyperkalemia (K+>5.3mmol/l)	54 / 101	53.5% [43.7;63.2]
VKA overdose (INR>4.9)	5/9	55.6% [23.1;88.0]
Renal failure (creat.>135 µmol/l or urea>16.6 mmol/l)	35 / 75	46.7% [35.4;58.0]
Other outcomes	14 / 53	26.4% [14.5;38.3]
TOTAL	108 / 238	45.4% [39.1;51.7]

2020-04-17

Pr Emmanuel Chazard - Artificial intelligence, Data reuse, Big data in Healthcare

How to evaluate an ADE detection rule?

- Evaluation of the accuracy of:
 - thiazide diuretic & renal failure → hyperkalemia
 - Which means: if the 3 conditions are present, then this is an ADE
- Several evaluations:
 - 1. The 2 causes are present, the outcome is present, and time consistent
 - 2. The 2 causes may explain the outcome
 - 3. The 2 causes are the main causes of the outcome
 - 4. There is a medication error
 - 5. There is a preventable medication error
- And: usability, acceptability, human factors ?





The ADE Scorecards, a webbased tool for ADE detection and visualization



The "ADE Scorecards", a tool for automated ADE detection in EHR







Scorecards / Tableaux de Bord

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Hospital / Hopital

Department / Service

Password / Mot de passe

Connection / Connexion



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Fungal infection (detected by prescription of a systemic antifungal)



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Number of stays with adverse events

Deselect all Select all	Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec
Anemia (Hb<10g/dl)	2	3	4	4	3	2	2	3	6	5	0	
Biological pancreatitis (amylase>90 UM or lipase>90 UM)	0	0	1	1	1	1	1	0	1	0	0	
Fungal infection (detected by prescription of a systemic antifungal)	5	3	2	2	2	0	1	1	3	2	0	
Hemorrhage (detected by a prescription of hemostatic)	2	0	2	2	4	5	4	2	2	2	2	
Hepatic cytolysis (alanine transa.>110 or aspartate transa.>110)	0	1	1	3	1	1	4	0	4	0	0	
High a CPK rate (CPK>195 UM)	0	0	0	0	4	1	0	0	0	0	0	
Hyperkalemia (K+>5.3)	9	10	12	8	11	15	7	10	11	6	0	
Hyponatremia (Na+<130)	1	0	0	2	1	0	0	0	1	0	0	
Mycosis (detected by the prescription of local antifungal)	0	0	0	0	0	1	0	0	0	0	0	
Neutropenia (count<1500/mm3)	0	0	1	0	0	0	0	0	1	0	0	
Renal failure (creat.>135 micromol/L or urea>16.6 mmol/L)	10	11	13	10	16	14	8	8	10	11	0	
Thrombopenia (count<75,000)	0	0	0	0	0	0	0	1	0	0	0	
Too high INR>4.9)	0	1	0	0	2	1	2	0	0	0	0	





Disconnect 🔞

Hyperkalemia (K+>5.3)

Characteristics of identified stays, all the rules together

Data from the coded diagnostics

Effective	65
Average age	75
Men - Women	42% - 58%
Deaths	15%

Number of cases per month



Cancers	25%
Cardiomyopathies	60%
Renal insufficiency	20%
Hepatic insufficiency	8%
Respiratory insufficiency	31%
Alcoholism	6%

Histogram of appearance delay



Conditions leading to Hyperkalemia (K+>5.3)

Number of cases

confidence ; median delay

(1) HMWH can induce hyperkaliemia, specially with renal insufficiency.

Second Kellinger A. D. Barlander Solid Incomentaria

<u>5</u>



Conditions leading to Hyperkalemia (K+>5.3)	Number of cases confidence ; median delay
 LMWH can induce hyperkaliemia, specially with renal insufficiency. 	2
Renal failure & Low weight heparin & Age < 70	17% ; 4.5j
(2) HMWH can induce hyperkaliemia, specially with ACE inhibitors and renal insufficiency.	2
Renal failure & High weight heparin & Angiotensin conversion enzyme inhibitor	13% ; 2j
(3) The suspension of some laxatives may reveal hyperkalemia.	<u>1</u>
Renal failure & Suspension of other laxative & Hepatic cholestasis	34% ; 2j
(4) The suspension of propulsive laxative may reveal hyperkalemia.	<u>1</u>
Renal failure & Propulsive laxative	34% ; 2j
(5) Angiotensin-converting enzyme and sartans may cause hyperkalemia.	<u>3</u>
Renal failure & Angiotensin conversion enzyme inhibitor & Opioïd	11% ; 2j
(6) The suspension of potassium lowering diuretic may reveal hyperkalemia.	<u>6</u>
Renal failure & Suspension of potassium lowering diuretic & NO Thrombin inhibitor	11% ; 3j
(7) The suspension of aminoglycoside may reveal hyperkalemia.	<u>1</u>
Renal failure & Suspension of aminoglycoside	17% ; 15j
(8) Prescription of nonsteroidal anti-inflammatory drugs may cause hyperkalemia.	1
Renal failure & NSAI & NO Potassium sparing diuretic	50% ; 1j
GLOBAL	. 65

<u>Confidence (a%)</u>: percentage of stays for which the effect occurs among the stays meeting the conditions. <u>Median delay</u>: from the moment when all conditions of the rule are met, period from which over 50% of effects will be appeared.



[1] HMWH can induce hyperkaliemia, specially with renal insufficiency.

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Details of rules

[1] HMWH can induce hyperkaliemia, specially with renal insufficiency.

Renal failure & High weight heparin \rightarrow Hyperkalemia (K+>5.3)

Somes aldosteronism or metabolic acidosis cases have been described with heparins. The risk is increased in case of a kidney insufficiency.

Ref. : Martindale - The complete drug reference- 34ème ed, The Pharmaceutical Press, London 2005 : 927-31.

In case of a high molecular weight heparin treatment, the dosage has to be adapted and the clinical and biological monitoring have to be increased.

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[2] HMWH can induce hyperkaliemia, specially with diabetic patients and renal insufficiency.

Renal failure & High weight heparin & Diabetes \rightarrow Hyperkalemia (K+>5.3)

Somes aldosteronism or metabolic acidosis cases have been described with heparins. The risk is increased in case of diabetes and renal insufficiency.

Ref. : Martindale - The complete drug reference- 34ème ed, The Pharmaceutical Press, London 2005 : 927-31.

In case of an high molecular weight heparin treatment, the dosage has to be adapted and the clinical and biological monitoring have to be increased.

С

[3] HMWH can induce hyperkaliemia, specially with ACE inhibitors and renal insufficiency.

Renal failure & High weight heparin & Angiotensin conversion enzyme inhibitor \rightarrow Hyperkalemia (K+>5.3)

Somes aldosteronism or metabolic acidosis cases have been described with heparins. The risk is increased in case of an angiotensin-converting enzyme inhibitor treatment and renal insufficiency.



Conditions leading to Hype	erkalemia (K+>5.3) Number of cases confidence ; median delay
 (1) LMWH can induce hyperkaliemia, specinsufficiency. Renal failure & Low weight heparin & Age < 70 (2) HMWH can induce hyperkaliemia, speciand renal insufficiency. Renal failure & High weight heparin & Angiotensing (3) The suspension of some laxatives matrix 	cially with renal 2 17% ; 4.5j cially with ACE inhibitors 2 conversion enzyme inhibitor 13% ; 2j Recrowark planning 1
 Renal failure & Suspension of other laxative & Her (4) The suspension of propulsive laxative Renal failure & Propulsive laxative (5) Angiotensin-converting enzyme and shyperkalemia. Renal failure & Angiotensin conversion enzyme in (6) The suspension of potassium lowering hyperkalemia. Renal failure & Suspension of potassium lowering (7) The suspension of aminoglycoside m Renal failure & Suspension of aminoglycoside (8) Prescription of nonsteroidal anti-infla hyperkalemia. Renal failure & Suspension of aminoglycoside 	Angiotensin-converting enzyme and sartans may cause hyperkalemia. July 2010 (0 cases): June 2010 (1 cases):
Confidence (a%): percentage of stays for which the effect occurs amor <u>Median delay</u> : from the moment when all conditions of the rule are met Details of rules	GLOBAL 65 ang the stays meeting the conditions. et, period from which over 50% of effects will be appeared.

[1] HMWH can induce hyperkaliemia, specially with renal insufficiency.

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Back to the stay	More information Rule info			PSIP
ATC 🔺	Drug name 🔶	0	5	10 🜩
	1000 mI POLYIONIQUE G5 + POTASSIUM CHL A + SPASFON SOL INJ	•		
02BC02	INIPOMP 20 MG, CPR (VOIR INEXIUM 20)			
06AD15	FORLAX 10G PDR ORALE SACHET PR SOL BUV			•
07DA03	IMODIUM 2MG GELULE VERT FONCE ET VERT CLAIR			
07XA04	TIORFAN 100MG GELULE BLEU CLAIR ET BLEU FONCE			
10AB05	NOVOMIX 30 FLEXPEN 100 U/ML, SUSP INJ, STYLO 3 ML			
12BA01	KALEORID 1000MG LP CPR ENR			
01AC06	KARDEGIC 75 MG, PDR PR SOL BUV, SACHET			
01DX12	CORVASAL 4MG CPR SECABLE BLANC			M
07AB03	ATENOLOL 50 MG ARROW, CPR			
09BA04	PRETERAX, CPR			
X10AA03	ELISOR 20MG CPR SECABLE = PRAVASTATINE 20 MG SANDOZ			
01DA23	OROKEN 200MG CPR PELLICULE (PRESCRIPTION PAPIER OBLIGATOIRE ANTIBIO)			
106DX02	TANAKAN 40MG CPR ENR BRUN ROUGE			
/03AE01	KAYEXALATE 454G PDR PR SUSP BUV/RECT			





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Back to the stay	More information Rule info	PSIP	Show all	Hide all INR V K1 L%	PROTIDES TOTAU Proteinurie RA.	UGM Glycosurie
ATC 🔺	Drug name 🔶	0 5 10 🜩	CL1	MONOCYTES%	TCA TCA TCA TCA TCA	🔲 Mo
	1000 mI POLYIONIQUE G5 + POTASSIUM CHL A + SPASFON SOL INJ	• I I I	CRP GB	MYELOCYTES%	TCAr ТСМН	Pb Pe
A02BC02	INIPOMP 20 MG, CPR (VOIR INEXIUM 20)		🗌 GGT.	- PB%	TGO.	🔲 Pn
A06AD15			(Millions/ml	PHAL.	П төг.	
A07DA03	Stay info - Mozilla Firefox		у н а нт	PLAQ.	UREE SG	
A07XA04	T http://www.expert-explorer.eu/stay_frames_info_rule.php?h	keffect_scoi 닸				
A10AB05	Ւ Rule info		1			
A12BA01	k hvnerkalemia (K+⇒5 3)					
B01AC06	HMG CoA reductase inhibitor & NO hypoalbuminemia & inflamm	nation → hyperkalemia (K+>5.3)			<u> </u>	
C01DX12	$_{ m C}$ NO renal failure & beta blocker $ ightarrow$ hyperkalemia (K+>5.3)		00			
C07AB03	NO renal failure & angiotensin conversion enzyme inhibitor \rightarrow h	nyperkalemia (K+>5.3)				
C09BA04	NO renal failure & potassium \rightarrow hyperkalemia (K+>5.3)				A COL	
C10000			0			
CTUAAU3	g Terminé	io 🖪			•	
J01DA23	OROKEN 200MG CPR PELLICULE (PRESCRIPTION PAPIER OBLIGATOIRE ANTIBIO)		3,00			
N06DX02	TANAKAN 40MG CPR ENR BRUN ROUGE		Ľ			
V03AE01	KAYEXALATE 454G PDR PR SUSP BUV/RECT		0	2	4 6	ō

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Back to the stay	Мо	re information Rule info	Show all	Hide all		PROTIDES	VGM
		Sans titre - Bloc-notes	DT.			Proteinurie RA.	Glycosurie
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	1000	cher confrere,		^	FES%	TCA patient	Mye
	SPAS	nous laissons sortir ce jour, @ @ @ @, nee le @/@/@, @ @ @ @ @ @ @, dans notre service du @/@/@ au @/@/@ pour diarrhees et vomissements.	hospitalise	e		ТСАГ	PD Pe
A02BC02	INIPO	antecedents				🔲 төо.	Pn
A06AD15	FORL	sur le plan chirurgical				TGP.	
A07DA03	IMOD	-cholecystectomie -prothese totale de hanche gauche sur coxarthrose		=		UREE SG	
A07XA04	TIOR	-prolapsus en © avec pose de pessaire -hysterectomie					
A10AB05	NOV	sur le plan medical -drid insulino-requerant					
A12BA01	KALE	-rgo -hta					
B01AC06	KARE	-dyslipidemie -cardiopathie ischemique				•	\sim
C01DX12	COR	-nernie uiscale avec cruralgie.				/	
C07AB03	ATEN						
C09BA04	PRET	histoire de la maladie la patiente a ete adressee aux urgences du centre 0 de 0 nour intole	rance				
C10AA03	ELIS(SANE	alimentaire dans un contexte de diarrhees, vomissements et febricule depuis le @/@/@.	a 38° evol	uant	<u> </u>	•	
J01DA23	ORO	examen clinique a son arrivee dans le service, la patiente presentait un etat genera	l relativer	ient	-		
N06DX02		conserve avec cependant une astnenie importante evoluant de facon ch presentait egalement une douleur lombaire chronique. il previstait pas d'amaignissement recent la patiente etait apyreti	nonique, el que	ie			
140000702	10190	sur le plan cardiovasculaire, les constantes etaient honnes avec une	tension			4 (5
V03AE01	KAYE	arterielle a 10/0 et une frequence cardiaque a 95. il n'existait pas fonctionnel cardiaque, a l'examen, on ne notait aucun signe d'insuff	de signe isance carc	liaque			
		droite ou gauche. l'auscultation etait sans particularite avec des bruits du coeur reg de souffle cardiaque ou vasculaire.	juliers, abs	ence			
		sur le plan respiratoire, saturation a 95% en air ambiant, la patien plaignait d'aucun signe fonctionnel respiratoire. l'auscultation eta particularite, les murmures vesiculaires etaient bilateraux et symet bruit surajoute.	te ne se it sans riques sans	×			10
			<				>

What did we learn about ADEs with the ADE Scorecards?

Installed in 5 hospitals (2 Danish, 2 French and 1 Bulgarian)

Routinely used by the physicians and pharmacists of a French general hospital during three years



Statistics about ADEs

	Nb of cases of outcome *		Nb of cases occurring during the stay		Nb of potential ADEs (automated detection)		Nb of confirmed ADEs (expert review)**	
Hyperkalemia	1301 2.67%	l.	703 2.84%	I	507 2.05%	-	271 1.1%	I
Renal failure	2293 <i>4</i> .7%		728 2.94%	I.	404 1.63%	I	189 <i>0.76%</i>	
VKA overdose	625 1.28%		321 <i>1.3%</i>	I	246 <i>0.99%</i>	_	137 <i>0.55%</i>	
Other kinds of outcomes	13936 28.56%		7171 28.97%		1438 5.81%		380 1.53%	I.
Total	14454 (29.62%)	%) 7624 (30.8%) 2196 (8.87		6 (8.87%)	997	(4.03%)	

*: the number of events is reported to the total number of stays, but next numbers are related to inpatient stays lasting at least 2 days

**: extrapolated from a sample

	Location	Hyperkalemia	Renal failure	VKA overdose	Other kinds of outcomes	Total
	All Hospitals	1.67%	1.62%	0.77% 📗	4.18%	7.23%
					A	
	Hospital	Hyperkalemia	Renal failure	VKA overdose	Other kinds of outcomes	l otal
	Hospital 1 (Dk)	0.47%	0.97%	0.10%	5.86%	6.78%
	Hospital 2 (Fr)	3.05%	1.53%	0.76% 📗	0.76% 📗	4.58%
E	Hospital 3 (Dk)	0.88% 📗	0.61% 📗	0.00%	4.02%	5.07%
	Hospital 4 (Fr)	10.57%	2.94%	0.00%	8.22%	18.20%
	Hospital 5 (Bu)	0.05%	0.02%	0.00%	0.14%	0.19%
· ′	Hospital 6 (Fr)	1.67%	1.62%	0.77% 📕	4.18%	7.23%
	Department of Hosp 6	Hyperkalemia	Renal failure	VKA overdose	Other kinds of outcomes	Total
,	Surgery	0.89% 📗	0.48% 📗	0.14%	3.09%	4.12%
1/2	Geriactrics	3.41%	3.41%	0.00%	5.40%	10.51%
K	Gyneco obst.	0.00%	0.00%	0.00%	0.41%	0.41% 📗
K	Cardiology	2.65%	2.79%	1.36%	4.23%	9.89%
	Internal medicine	2.67%	2.58%	1.62%	10.22%	14.61%
,	Pneumology	2.77%	2.66%	2.11%	6.21%	11.86%
	Department of Hosp 1	Hyperkalemia	Renal failure	VKA overdose	Other kinds of outcomes	Total
X	Neurology	1.09%	2.46%	0.82%	6.83%	9.84%
	Cardio/endoc.	1.17% 📗	2.09%	0.31% 📘	6.28%	8.38%
	Geriatrics	1.02%	2.24%	0.20%	17.48%	18.70%
	Gynecology	0.00%	0.00%	0.00%	0.00%	0.00%
K	Intensive care	0.28%	1.40%	0.00%	11.76%	13.17%
	Internal Med.	0.73% 📗	1.52%	0.00%	10.46%	11.92%
	Obstretrics	0.00%	0.00%	0.00%	1.47%	1.47%
	Orthopedic	0.18%	0.80% 📗	0.00%	8.57%	9.10%
	Rheumatology	1.12%	1.57%	0.22%	18.61%	20.18%
	Urology	0.36%	0.72%	0.09%	4.43%	5.15%

Characteristics of the patients (n=73,836 inpatient stays)

Parameter	Overall	Cardiology	Gyn. Obs.	p value (all)
Age (years)	60.2	2 67.6	5 28	p<0.001
Length of stay (days)	8.0 2	1 8.19) 11.6	p<0.001
Proportion of men	40.80%	42.80 %	6.00%	p<0.001





Surgery
Cardiology
Internal medicine
Pneumology
Gynecology Obstetrics
Geriatrics



Potential ADE cases with INR increase (INR≥5)

due to vitamin K antagonist overdose or interaction may induce a severe hemorrhage

Estimated proportion:

0.99% [0.89%;1.09%]

Interesting data:



Parameter	Overall	Cardiology	Gyn. Obs.	p val (all)
VKA	8.34%	15.50%	0.00%	p<0.001
Chr. hepatic insuf.	4.90%	13.70%	0.04%	p<0.001
INR increase (all)	2.46%			
Potential ADE	0.99%	1.36%	0.00%	p<0.001

Potential ADE cases with hyperkalemia (K⁺>5.5mmol/l)

may induce lethal cardiac rhythm troubles

Estimated proportion:

2.03% [1.89%;2.18%]

Interesting data:



Parameter	Overall	Cardiology	Gyn. Obs.	p val (all)
Diuretics	23.30%	41.10%	0.00%	p<0.001
Chr. renal insuf.	2.02%	3.04%	0.04%	p<0.001
Hyperkalemia (all)	5.43%			
Potential ADE	2.03%	2.65%	0.00%	p<0.001

For each ADE detection rule, contextualized statistics are computed in each setting

Rule: vitamin K antagonist & amoxicilline&clav.ac. & age $\geq 70 \rightarrow$ appearance of high INR (INR ≥ 5)

Department		Confidence (PP	V) Support (frequency)	Median delay			alue
X all department	nts	10/57=17.5	5% 10/5322=1.9%	6.5j	13.38	0	
Department	Confidence	(PPV) Sup	pport (frequency)	Median o	lelay Rel	ative risk I	isher's test P value
X all dpts	10/57=3	17.5%	10/5322= 1.9‰	6.5j		13.38	0
X medicine B		3/17=17.7	7% 3/966= 3.1%	Зј	9.31	0.005	
X pneumology		5/28=17.9	9% 5/818=6.1%o	11j	6.41	0.0016	
Y all departmer	nts	1/10=10)% 1/11923= 0.1%	6j	33.09	0.0306	
Y apoplexy		No stay matches	s the conditions				
Y cardio & end	ocrinology	1/2=50)% 1/1967= 0.5%	6j	51.71	0.0202	
Y geriatrics		0/2=0	0/493=0%0		0	1	
Y gynecology		No stay matches	s the conditions				
		No stay matches	s the conditions				
Y internal medi		0/5=0	0% 0/1514=0%		0	1	
Y obstetrics		No stay matches	s the conditions				
Y orthopedics		No stay matches	s the conditions				
Y rheumatology		No stay matches	s the conditions				
Y urology		No stay matches	s the conditions				
Z all departmer	its	0/1=0	0% 0/1022=0%		0	1	
W all departme	nts	0/8=0	0/7685= 0% 0		0	1	

Comment:

- Les pénicillines augmentent le risque hémorragique sous AVK.
 Une augmentation de l'activité des anti-vitamine K peut être observée chez les patients traités par pénicillines. Le risque hémorragique est accru Ref : thésaurus IAM-AFSSAPS juin 2009.
- En cas de prise de pénicillines, la posologie des AVK sera adaptée et la surveillance clinique et biologique accrue.

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Toward an ADE prevention by CDSS based on level 2 artificial intelligence



Level 2 AI: the failure of fullyautomated machine learning

Supervised data mining:

- Good predictive power
- Enables to filter, reorganize and explain knowledge
- "Black boxes", such as deep learning
 - Better predictive power
 - Does not enable to manage knowledge!

However, some other steps are from far more crucial: feature extraction

Prospective prevention of ADEs



Alert generation, before the ADE occurs, in order to prevent it.

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Prospective prevention of ADEs Our contextualized approach

- E.g. VKA & PPI \rightarrow risk of hemorrhage
- Usual implementation of alerts:

Medical unit A	Medical unit B	Medical unit C	
VKA & PPI → interruptive alert			

PSIP's contextualized implementation of alerts:

Medical unit A	Medical unit B	Medical unit C		
Empirical probability=10%	Empirical probability=0.01%	Unseen circumstances		
VKA&PPI→	VKA&PPI→	VKA&PPI→		
interruptive alert	silent or non-interruptive alert	interruptive alert		

... and "personalized medicine"!

Unless age>70

Prospective prevention of ADEs The PSIP approach

- Implementation of 3 CDSS:
 - The IBM prototype
 - The Medasys prototype
 - The PSIP prescription simulation
- Major characteristics:
 - Filtering of alerts, based on contextualized statistics
 Jess alerts, more accurate
 - More complex rules (statistical segmentation)
 more accurate evaluation of the probabilities
 - Innovative alert methods:
 - less interruptive
 - Showing actual pas ADE cases => more acceptable



Benefits expected from the computation of empirical risk



2020-04-17

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What is the empirical risk of a drug administration?



Quantitative studies in CDSS evaluation



Quantitative studies in CDSS evaluation – our approach



Automated computation of the probability of outcome for each DDI rule



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Results for INR ≥ 5

- Drugs with Hazard ratio significantly \neq 1:
- Fenofibrate HR=3.09
 [1.34; 7.13]
- Methylprednisolone HR=3.02 [1.37; 6.68]
- Simvastatin HR=2.52 [1.36; 4.67]
- Ofloxacin HR=0 n=123

	Drug category	Drug interacting with VKA	Outcome	Number	HR	p value
	Analgesics and	Etodolac	INR≤1.5	3	0	0.72
	antipyretics	Paracetamol (acetaminophen)	INR≥5	958	0.94	0.8
		Tramadol	INR≥5	431	0.65	0.22
	Antibiotics	Amoxicillin	INR≥5	781	1.12	0.65
		Amoxicillin and enzyme inhibitor	INR≥5	666	1.24	0.4
		Ciprofloxacin	INR≥5	124	1.14	0.83
		Clarithromycin	INR≥5	29	0.96	0.97
		Cloxacillin	INR≤1.5	14	0	0.17
		Dicloxacillin	INR≤1.5	24	4.58	0.04 ***
		Erythromycin	INR≥5	5	8.43	0.13
		Levofloxacin	INR≥5	44	2.04	0.37
		Metronidazole	INR≥5	83	2.06	0.21
		Norfloxacin	INR≥5	18	0	0.31
		Ofloxacin	INR≥5	123	0	0.009 ***
		Rifampicin	INR≤1.5	30	0.55	0.51
		Teicoplanin	INR≤1.5	30	0.89	0.91
		Tetracycline	INR≥5	7	0	0.56
		Tranexamic acid	INR≥5	14	0	0.51
		Trimethoprim sulfamethoxazole	INR≥5	31	0	0.18
	Cardiovascular &	Amiodarone	INR≥5	670	0.94	0.81
	anti-hypertensive	Candesartan	INR≤1.5	164	1.24	0.57
	urugs	Diltiazem	INR≥5	173	1.23	0.63
		Disopyramide	INR≥5	5	0	0.85
		Furosemide	INR≤1.5	1551	0.89	0.45
		Propafenone	INR≥5	4	0	0.75
		Propranolol	INR≥5	41	0	0.26
		Telmisartan	INR≤1.5	15	1.22	0.85
	Central nervous	Barbiturates	INR≤1.5	15	1.63	0.53
	system drugs	Carbamazepine	INR≤1.5	13	1.78	0.46
		Chlordiazepoxide	INR≤1.5	5	0	0.88
		Citalopram	INR≥5	79	0.95	0.93
		Fluoxetine	INR≥5	26	1.63	0.65
		Quetiapine	INR≥5	5	5.63	0.18
		Ropinirole	INR≥5	3	0	0.67
		Sertraline	INR≥5	21	1.85	0.58
	Drugs acting on	Acetylsalycilic acid	INR≥5	683	1.04	0.87
	hemostasis	Heparin (unfractionated)	INR≥5	294	0.6:	0.35 😑
	Drugs for acid-	Omeprazole	INR≥5	130	0.32	217
llia	related disorders	Sucralfate	INR≤1.5	11	9	0.005 ***
이 이 주말						

Results for INR ≤ 1.5

- Drugs with Hazard ratio significantly \neq 1:
- Dicloxacillin HR=4.58
 [1.81; 11.56]
- Sucralfate HR=9 [3.13; 25.88]

Drug category	Drug category Drug interacting with VKA		HR	p value
Analgesics and antipyretics	Etodolac	3	0	0.72
Antibiotics	Cloxacillin	14	0	0.17
	Dicloxacillin	24	4.58	0.04 ***
	Rifampicin	30	0.55	0.51
	Teicoplanin	30	0.89	0.91
Cardiovascular & anti-	Candesartan	164	1.24	0.57
hypertensive drugs	Furosemide	1551	0.89	0.45
	Telmisartan	15	1.22	0.85
Central nervous system drugs	Barbiturates	15	1.63	0.53
	Carbamazepine	13	1.78	0.46
	Chlordiazepoxide	5	0	0.88
Other drugs	Azathioprine	15	0	0.26
	Bosentan	5	0	0.49
	Chelation therapy	5	0	0.57
	Mesalazine (5-ASA)	10	0	0.31
	Sucralfate	11	9	0.005 ***
	Sulfasalazine	8	3.23	0.33



Apparent discrepancy between our results and the literature

Outcome	Drug	Causation (review from ³⁷)	Severity (review from ³⁷)
	Fenofibrate (HR>1)	Highly probable	Moderate
	Methylprednisolone (HR>1)	Highly improbable	Major
	Simvastatin (HR>1)	Probable	Minor
INR≥5	Ofloxacin (with HR<1)	Possible	Major
		highly improbable: 9%	nonclinical: 50%
	85 other drugs	possible: 32%	minor: 4%
	(HR not different from 1)	probable: 30%	moderate: 33%
		highly probable: 29%	major: 13%
	Sucralfate (HR>1)	Highly probable	Non clinical
	Dicloxacillin (HR>1)	Probable	Moderate
INR<1.5		highly improbable: 14%	nonclinical: 46%
	31 other drugs	possible: 14%	minor: 7%
	(HR not different from 1)	probable: 29%	moderate: 36%
		highly probable: 43%	major: 11%

=> Empirical probabilities take into account the knowledge and monitoring of the physician, not only theoretical knowledge

Toward a new paradigm for ADE prevention?

- Mixed results of ADE prevention by alert systems:
 - Over-alerting, poor positive predictive value
 - Alert fatigue, poor alert acceptation, even when appropriate
 - Poor clinical impact (except dose computation in pediatrics)
- Hypothesis: let's calibrate CDSS on the empirical risk, and not the theoretical risk!



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We also learnt: finally, artificial intelligence should integrate the human workflow



Clinical evaluation of the ADE Scorecards



Clinical evaluation of the ADE Scorecards

support to a global quality improvement approach:



More generally speaking

- Digitization (1990-2015):
 - Notably driven by the problem of the cost of labor. Example of France:
 - Cost of labor = net salary * 1.85
 - 35 hours per week
 - Increase of support services vs business units
 - Perverse effects:
 - A part of the population becomes unemployable...
 - Work transfer from poorly qualified people to managers
 - And then, managers work below their qualification level
- A "good" artificial intelligence:
 - May respect the human workflow
 - May help transferring tasks to less qualified people
 - Based on current scientific knowledge, requires human validation



Thank you for your attention!

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