

## Extracting food-drug interactions from scientific literature

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#### Context



- Food-drug interaction  $\Longrightarrow$  Adverse effects
- Less known and sparse in unstructured data
   E.g.: food-drug interactions
   Grapefruit juice increases effect of other dihydropyridine calcium antagonists.

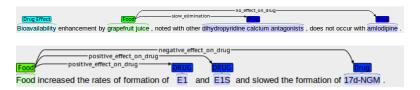
unlike for drug-drug interaction or drug adverse effect (DrugBank  $^{1}$  or Theriaque  $^{2}$ )

**Goal**: Automatic identification of interaction statements between drug and food in abstracts of scientific articles issued from the Medline database.

Approach: Use of NLP methods for scientific abstracts mining

#### **Problematic**





- Variable mention of drugs and foods in abstracts
- Fine description of interactions
- Unbalanced learning set
  - $\implies$  14 types of relation, 831 sentences

#### Corpora



639 Medline abstracts with the query

Brat annotation by an intern in pharmacy [Hamon et al.17]

Relation	#	# % Relation		#	%
unspecified relation	530	58,8%	no effect on drug	109	12,1%
decrease absorption	53	5,9%	improve drug effect	6	0,7%
positive effect on drug	21	2,3%	without food	13	1,4%
negative effect on drug	88	9,8%	speed up absorption	1	0,1%
increase absorption	39	4,3%	worsen drug effect	8	0,9%
slow elimination	15	1,7%	new side effect	4	0,4%
slow absorption	15	1,7%			
Total			100%		

## Grouping relation



- Intuitive grouping
- Unsupervised clustering
- Drug-drug interaction Domain adaptation

## Intuitive grouping (ARNP)

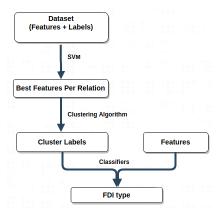


- Non-precised relation
- No effect
- Reduction decrease absorption, slow absorption, slow elimination
- Augmentation
   increase absorption, speed up absorption
- Negative new side effect, negative effect on drug, worsen drug effect, without food, negative effect on drug, worsen drug effect, side effect, without food
- Positive positive effect on drug, improve drug effect

### Relation Clustering



- Relation representation method
- Clustering method on types of relation



#### Results



- New grouping scheme: (1) decrease absorption, increase absorption, (2) improve drug effect, new side effect, worsen drug effect, which refer to effect of drug, speed up absorption, slow absorption, without food, positive effect on drug, (3) negative effect on drug, (4) no effect on drug, (5) slow elimination
- Improvement on F1-score with 200 features: from 0.41 with ARNP and non-clustered data to 0.58
- Reduction of the impact of the imbalance of data: Difference of macro and micro F1 from 0.23 to 0.09

## Domain adaptation - Drug-drug interaction



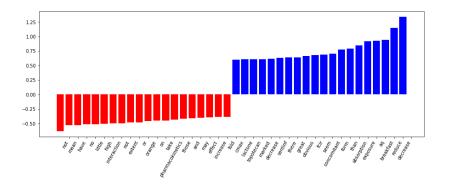
Correspondence of type DDI-FDI (Line 1) et percentage of FDI instances affected to type DDI (Lines 2-5)

Relation (Rel), Decrease absorption (Dec), No effect on drug (No), Increase absorption (Inc), Negative effect on drug (Neg), Positive effect on drug (Pos), New side effect (New), Without food (Wout), Improve drug effect (Imp), Slow elimination (SI-e), Slow absorption (SI-a), Worsen drug effect (Wors), Speed up absorption(Speed), Advice (A), Mecanism (M), Effect (E), Interaction (Int)

FDI	Rel	Dec	No	Inc	Neg	Pos	New	Wout	lmp	SI-e	Wors	SI-a	Speed
DDI	М	М	М	М	E	Е	E	Α	E	М	E	М	М
Advice	7	4	10	8	12	24	0	54	17	0	0	0	0
Effect	50	7	31	13	69	48	100	23	83	13	75	0	0
Int	15	0	0	0	7	0	0	0	0	0	0	0	0
Mecha	28	89	59	79	11	29	0	23	0	87	25	100	100

⇒ F1-score from 0.41 on the initial labels to 0.78 on the new labels

# 20 best and worst SVM features coefficient for Decrease absorption relation



#### Bibliographie

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