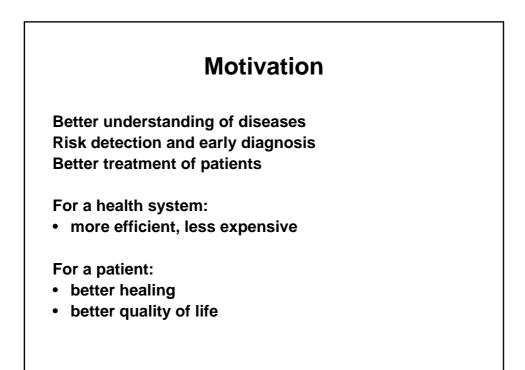
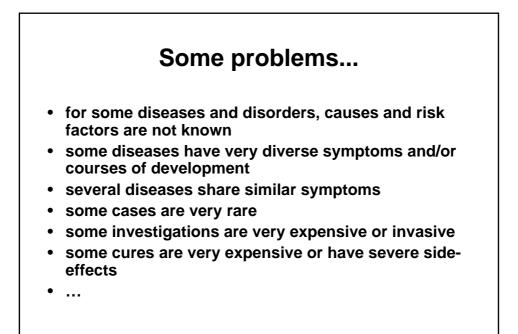
Intelligent Systems for Knowledge Discovery in Biomedical Field

Tanja Urbancic University of Nova Gorica, Nova Gorica, Slovenia and Jozef Stefan Institute, Ljubljana, Slovenia

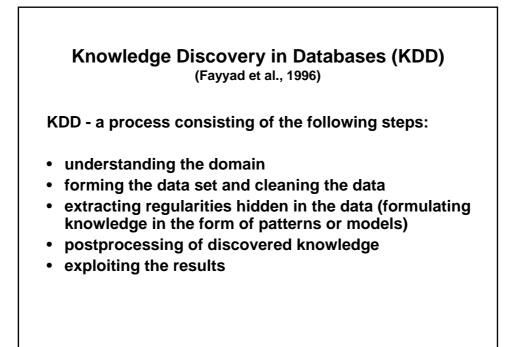
In cooperation with: Ingrid Petric Bojan Cestnik Marta Macedoni-Luksic

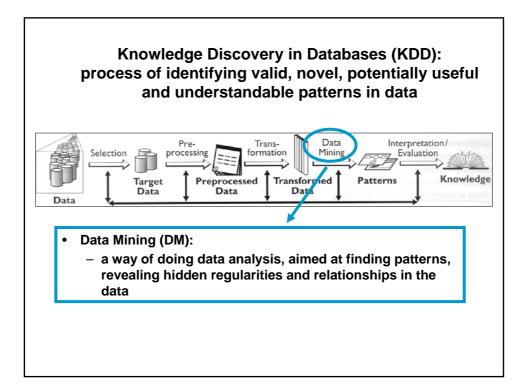
Summer School on Intelligent Systems, University of Cyprus, July 2-6, 2007









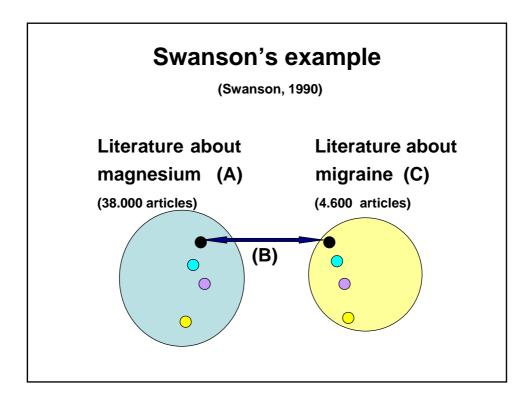


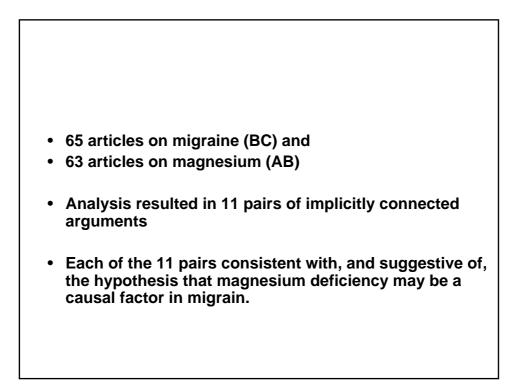
Data mining in medicine

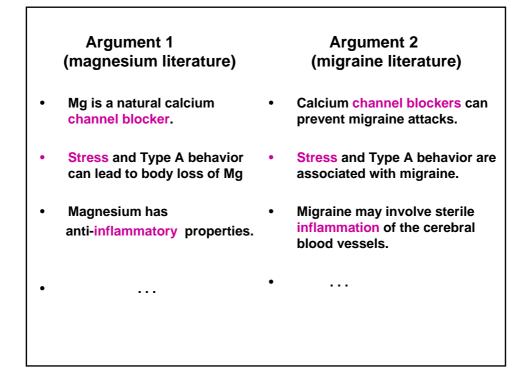
- Large quantities of data are collected
- Most often
 - predictive DM used for classification models (for diagnosis, prognosis, treatment planning)
 - data (represented in tables) collected from measurements or acquired by experts
 - overview of methods, examples and an exahaustive list of references e.g. in (Lavrac and Zupan, 2005)
- Data in different forms
 - e.g. images, texts

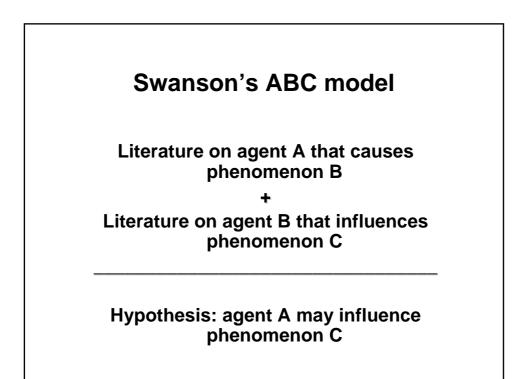
Text mining in biomedicine (overview and examples in Cohen and Hersh, 2005) Extracting interesting information from biomedical

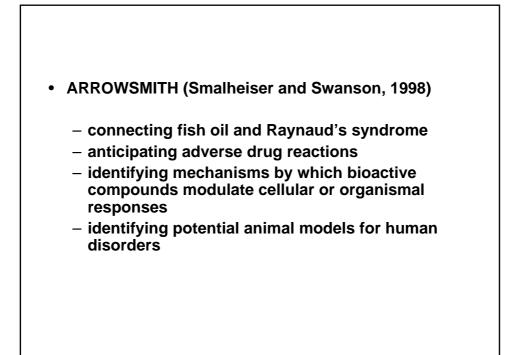
- Extracting interesting information from biomed knowledge represented in digital text forms
- Used also for
 - Relationship extraction (to recognize occurences of a pre-specified type of relationship, e.g. between genes and proteins)
 - Hypothesis generation (to uncover implicit relationships, worthy further investigation, e.g. potential new uses of drugs)

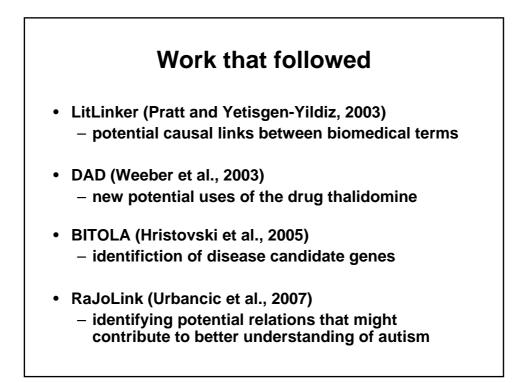












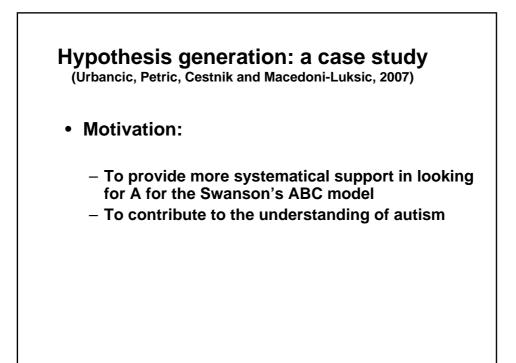
Generation of a hypothesis A may influence C

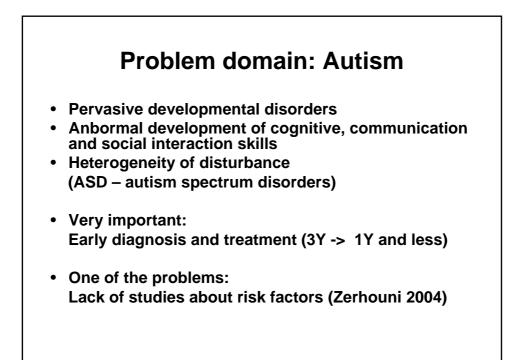
For a given C, how do we find A?

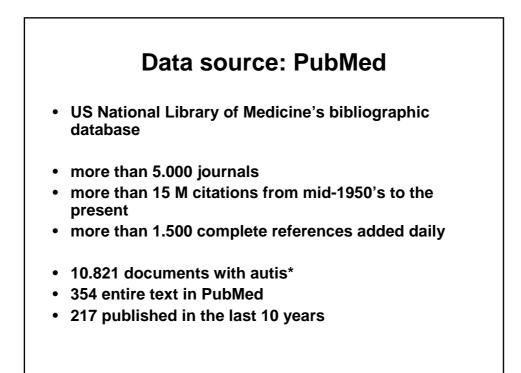
Swanson:

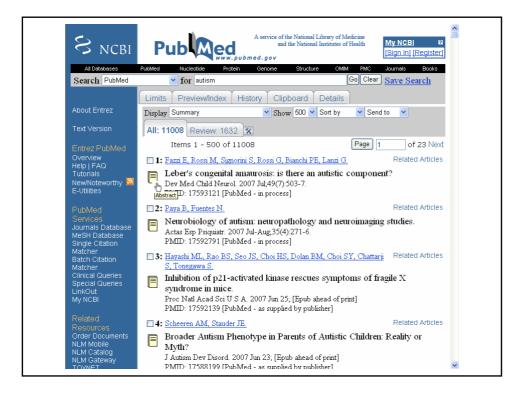
Search proceeds via some intermediate literature (B) toward an unknown destination A. ... Success depends entirely on the knowledge and ingenuity of the searcher.

Wheeler: Our whole problem is to make the mistakes as fast as possible.

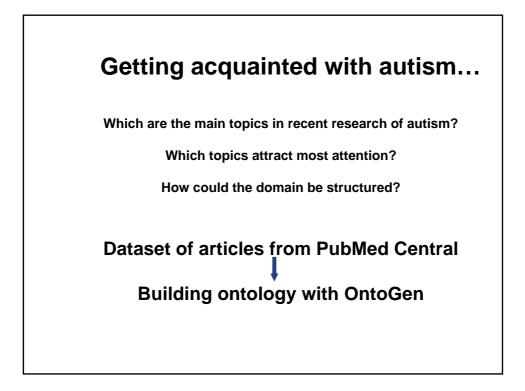


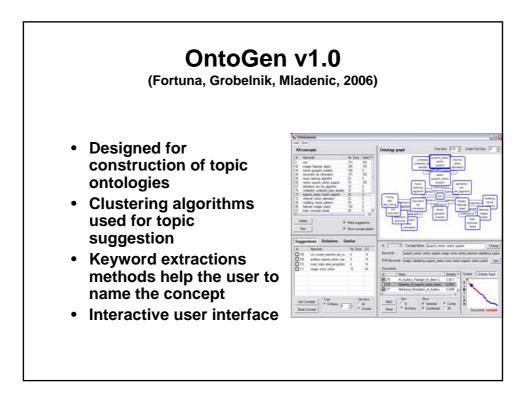


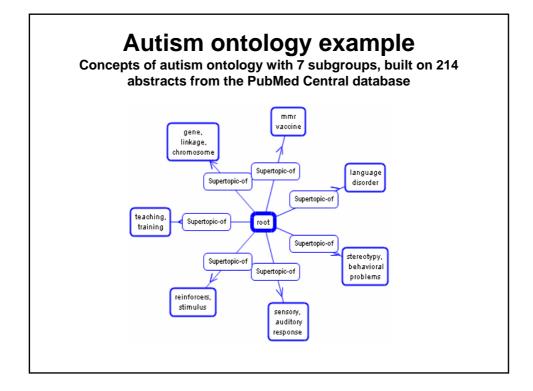


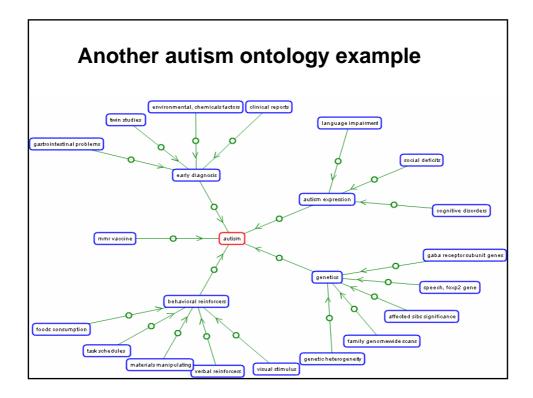


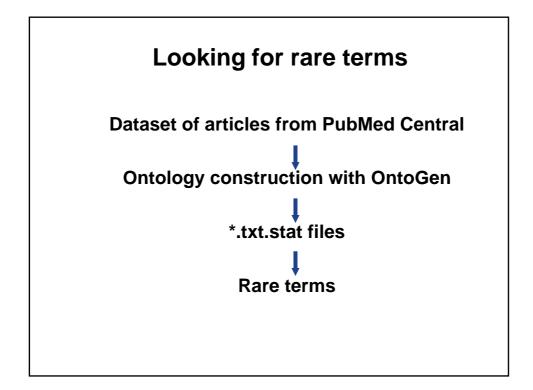












214Texts.txt.stat - Beležnica		
Datoteka Urejanje Oblika Pogled Pomoč		
TELEPHONED RADIO 113277.	1 'TEXT.BOX'013265. 1 'TEXTBOX'013266. 1 1 'TETRAHYDROBIOPTERIN'013270. 1 'TETS.SAM'013272. 1 1 'TETS.SAM'013272. 'TETS.SAM'013273. 1 1 1 'TETS.SAM'013272. 'TETS.SAM'013273. 1 1 1 'TEST.SAM'013272. 'TEST.SAM'013276. 1 1 1 'TEXT.SAM'013277. 'TENTING_TAL'013276. 1 1 1 'TEACHING_VOCAS D13278. TEACHING_TAL'013276. 1 1 1 TEACHING_VOCAS D13278. TEACHING_TAL'013283. 1 1 1 PREFER'013284. 'TARGETED_FORMER'013293. TAGETED_PROMPED'013297. 1	ED'0 1. 4. 1 1 0. 318. 324.
1 'SUBITIZING'013325.	1 'STUDY_PUNISHED'013326. 1 'STUDENT_DIRECT_INSTRUCT'013327.	
214Texts.txt.stat - Beležnica		
Datoteka Urejanje Oblika Pogl <u>e</u> d Pomoč		
LARSE 14439.1 - LARGE LARSE LAPSED UTDEDSONNGGRAPHY'014192. LARSED UTDEDSONNGGRAPHY'014192. LANGUAGE_SUBTYPES'014194. LAMINAR_COMPARTMENT'014197. 14200 LACTOYLGLUTATHI KRISTINA'014204 1 'KTAASGE0REY'014212. 1 'KALSCHEURE_AL'014218. 1 'KALSCHEURE_AL'014218. 1 - JULI'05'014222. 1 - JULI'05'014222. 1 - JULI'05'014222. 1 - JEFF'014230. 1 - JEFF'014234. 1 - JE	1 'LEARNING_OUTCOME' 114187. 1 'LD-HD' 114188. 1 1 'LANGUAGE_SUBTYPES_GROUPS' 114193. 1 1 . 1 . LANGUAGE_SUBTYPES_GROUPS' 114193. 1 . 1 . LANGUAGE_SUBTYPES_GROUPS' 114193. 1 . LANGUAGE_SUBTYPES_GROUPS' 114210. 1 . LANGUAGES. 1 . LAGS_SCHEDUEL' 114210. 1 . . KELSEY 114210. 1 . KELSEY 114210. 1 . <td>1 1 25. 9. 1</td>	1 1 25. 9. 1

For further investigation we choose:

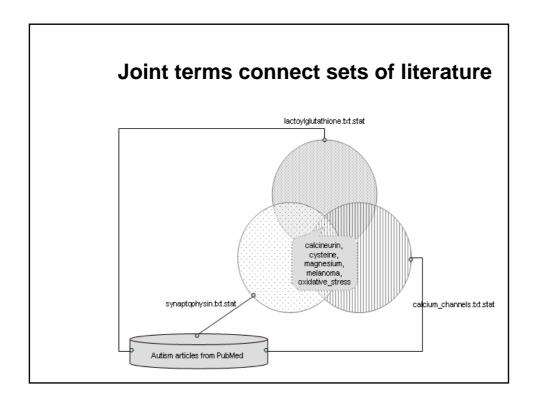
- lactoylglutathione
- synaptophysin
- calcium channels

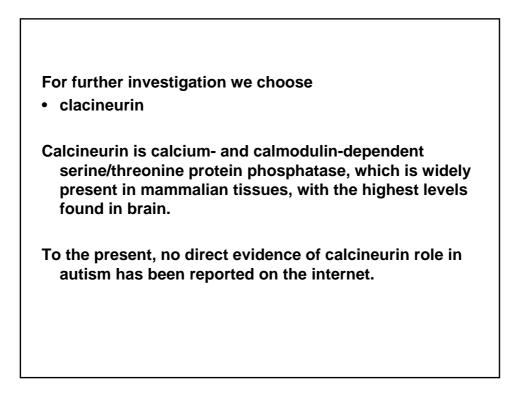
Why?

- icrease of polarity of glyoxalase I in autistic brain, glyoxalase system involves lactoylglutathione
- altered synaptic function in autism, synaptophysin is a protein localized to synaptic vesicles
- abnormal calcium signalling in some autistic children

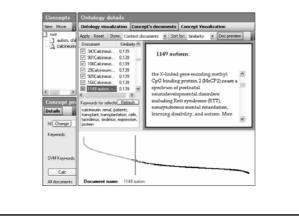
Do chosen rare terms have something in common?

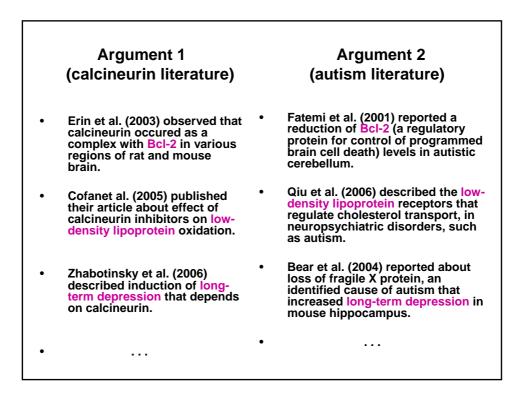
		ing	ı for joint	terms	
	Word	Total	calcium_channels	lactoylglutathione	synaptophysin
	'CYPRUS_INSTITUTE'	1	1		
	'CYS'	2	1		1
	'CYS_CYS'	1			1
	'CYS_CYS_CYS'	1			1
\mathbf{F}	'CYSTEINE'	3	1	1	1
	'CYSTEINE_MOTIF'	1			1
	'CYSTEINE_RESIDUE'	1	1		
	'CYSTEINE_RESIDUES'	2		1	1
	'CYSTEINE_RICH'	1	1		
	'CYSTEINE_RICH_DOMAINS	1	1		
	'CYSTEINE_STRING'	1			1
	'CYSTEINE_STRING_PROTE	1			1
	'CYSTEINYL'	1		1	
	'CYSTEINVLGLYCINE'	1		1	
Re	cord: 🚺 🖣 🚺 5654 🕨	► ►	of 30071		





OntoGen's representation of the set of autism and calcineurin articles according to their similarities. Two main topics (*autism and calcineurin*) are listed on the left side of the OntoGen's window. As the calcineurin is selected, the list of documents that are in the relationship with it is presented in the central part of the window. An outlying autism article (*1149 autism*) is inside the calcineurin context documents due to its similarity with the neighboring documents.





Towards the hypotheses... (1)

Which rare terms are promising for hypotheses generation?

Background knowledge is crucial.

Can we automatise selection of promising rare terms?

Partially, in some cases. E.g., selecting terms from a neurobiological dictionary.

But at the moment,

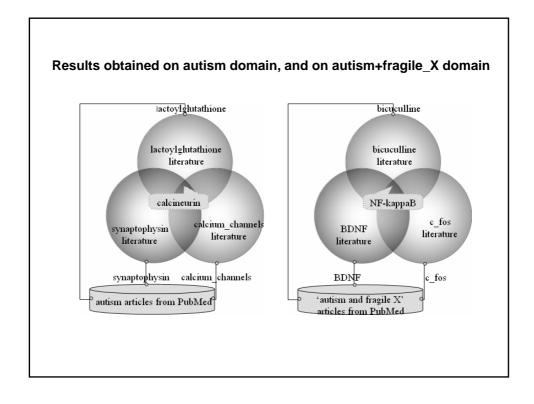
Expert's involvement is crucial.

Towards the hypotheses... (2)

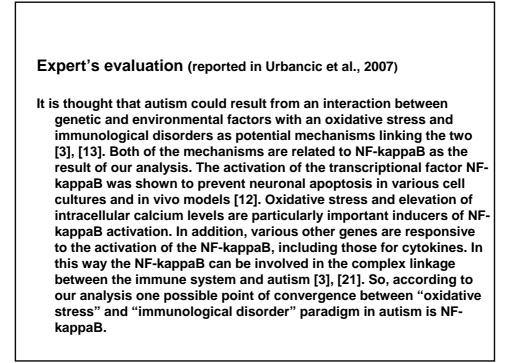
Do these pairs of documents point towards useful hypotheses?

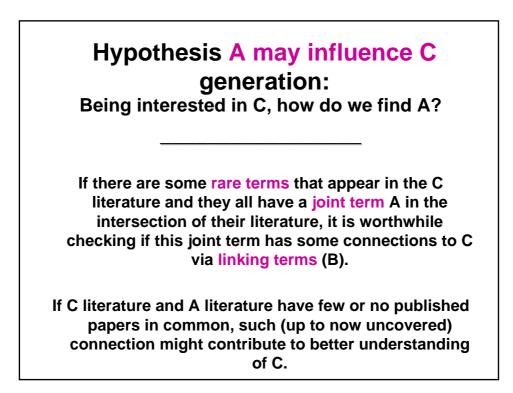
Expert's evaluation is crucial.

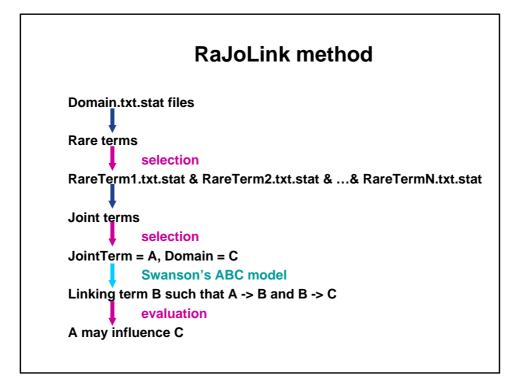
"Continue with fragil X, this would really be interesting..."



Autism literature	NF-kappaB literature
Araghi-Niknam and Fatemi [2] showed reduction of <i>Bcl-2</i> , an important marker of apoptosis, in frontal, parietal and cerebellar cortices of autistic individuals.	Mattson [12] reported in his review that activation of NF-kappaB in neurons can promote their survival by inducing the expression of genes encoding antiapoptotic proteins such as <i>Bcl-2</i> and the antioxidant enzyme Mn-superoxide dismutase.
Vargas et al. [21] reported altered cytokine expression profiles in brain tissues and cerebrospinal fluid of patients with autism.	activation NF-kappaB regulates the
increased urinary excretion of an	Zou and Crews [24] reported about increase in NF-kappaB DNA binding following oxidative stress neurotoxicity.







Step	Input	Action	Tool, technique	Human involvement	Output
Ra	Set of articles about domain of interest (about phenomenon C)	1.1 Extraction of texts	Digital document archives		
		1.2 Data collection preprocessing	Word processing software		
		1.3 Identification of rare (R) terms	Word frequency statistics		
		1.4 Semantic filtering	Latent semantic indexing	Indication of interesting R terms	R terms C_R ₁ C_R ₂ ,C_R _p

Step	Input	Action	Tool, technique	Human involvement	Output
Jo	Sets of articles about C_R ₁ , C_R ₂ ,, C_R _p	2.1 Extraction of texts	Digital document archives		
		2.2 Data collections preprocessing	Word processing software		
		2.3 Identification of each dataset's concepts and subconcepts	Word frequency statistics, Clustering		
		2.4 Search for joint terms	Word frequency statistics	Selection of significant joint terms	Joint terms A ₁ , A ₂ ,A _c (agents A)

Steps of the RaJoLink method (3)						
Step	Input	Action	Tool, technique	Human involv.	Output	
art	Joint set of articles about A _i	3.1 Extraction of texts	Digital document archives			
	and articles about C	3.2 Data collection preprocessing	Word processing software			
		3.3 Identification of semantically related A _i and C documents	Semantic text analysis			
		3.4 Search for linking terms (agents B)	Word intersection	Selection of meaningful terms B _i	Linking term s B ₁ B ₂ , B _r	

Case-study conclusions

- Ontology construction is useful for systematical exploration of sets of articles and for getting insight into a new domain.
- It is worthwhile to explore rare terms for generation of hypotheses (RaJoLink method).
- Expert's involvement is crucial for speeding up the process (selections) and for evaluations of candidate hypotheses.
- Expert evaluation confirmed the relevance of discovered relations in the autism domain.

Selected bibliography

- U.M. Fayyad, G. Piatetski-Shapiro, P. Smith: The KDD process for extracting useful knowledge from volumes of data. *Communications of the ACM*, 39(11), 27-41, 1996
- N. Lavrac, B. Zupan: Data Mining in Medicine. In Maimon and Rokach (eds.) *Data mining and knowledge discovery handbook.* New York: Springer, pp. 1107-1137, 2005
- M. van Sommeren, T. Urbancic: Applications of machine learning: matching problems to tasks and methods, *The Knowledge Engineering Review*, Vol. 20(4), 363-402, 2006
- A.M. Cohen, W.R. Hersh: A Survey of Current Work in Biomedical Text Mining. Briefings in Bioinformatics, 6(1), pp. 57-71, 2005.
- D.R. Swanson, Medical literature as a potential source of new knowledge. Bulletin of the Medical Library Association, 78(1), 29-37, 1990

Selected bibliography

(continued)

- T. Urbancic, I. Petric, B. Cestnik, M. Macedoni-Luksic: Literature Mining: Towards Better Understanding of Autism. In: *Artificial Intelligence in Medicine LNAI 4594* (R. Belazzi, A. Abu-Hanna, J. Hunter eds.), Springer, 215-224, 2007
- I. Petric, T. Urbancic, B. Cestnik: Discovering Hidden Knowledge from Biomedical Literature, *Informatica* 31, 15-20, 2007
- B. Fortuna, M. Grobelnik, D. Mladenic: Semi-automatic Data-driven Ontology Construction System. *Proceedings of the 9th international multi-conference Information Society*, Ljubljana, Slovenia, 223-226, 2006
- OntoGen: http://ontogen.ijs.si/
- PubMed: Overview at http://www.ncbi.nlm.nih.gov/