Biomedical and Health Informatics: From Foundations to Applications to Policy

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Detection of underspecifications in SNOMED CT concept definitions using language processing

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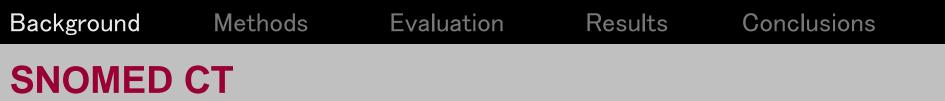
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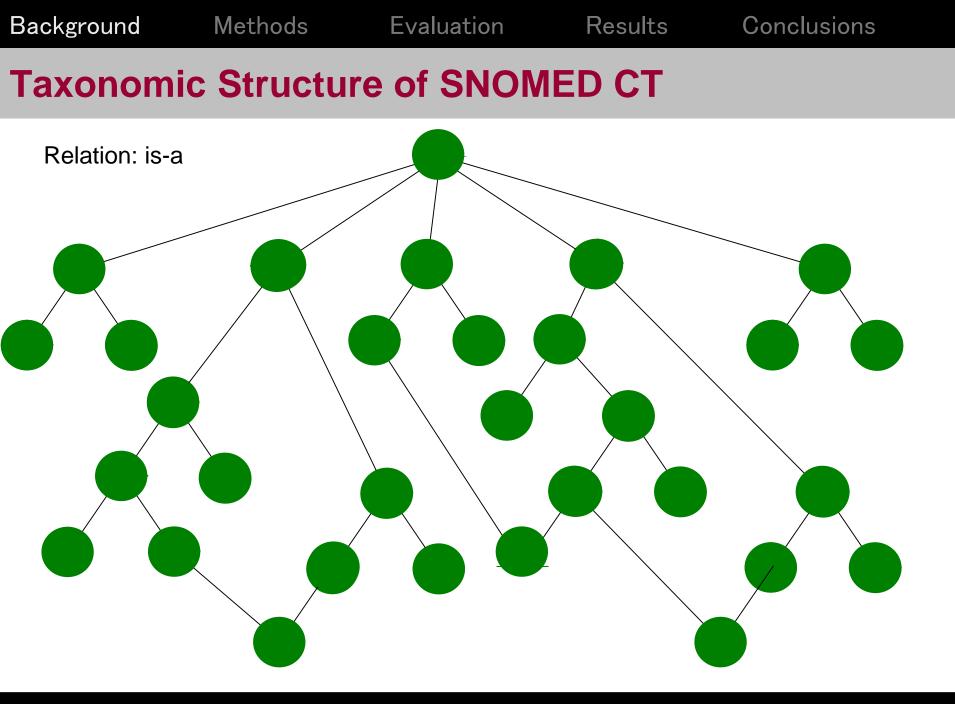






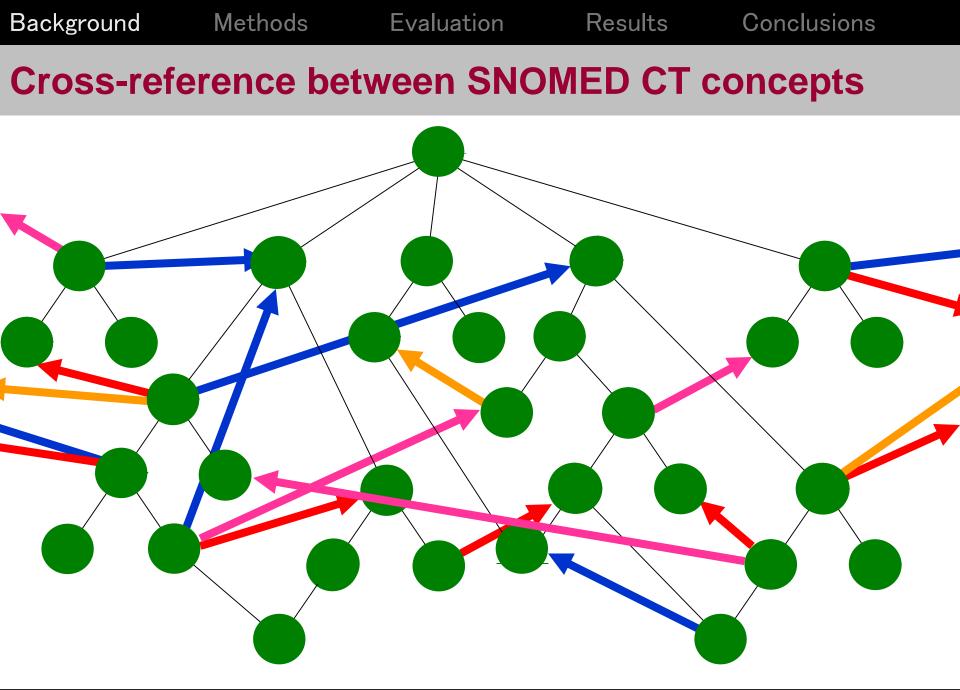


- "Standardized Nomenclature of Medicine Clinical Terms"
- Comprehensive clinical terminology (> 300,000 representational units)
- Concepts are arranged in extensive taxonomic (is-a) hierarchies



Background Methods Evaluation Results Conclusions SNOMEDCT

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- Cross-reference between concepts from several branches via semantic relations obeying description logics semantics



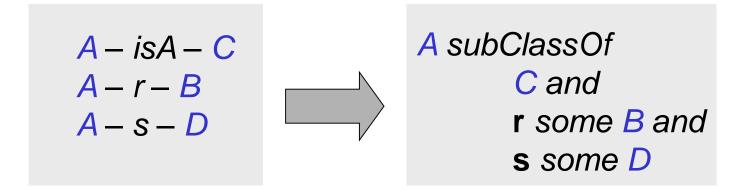
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 SNOMED CT semantics in a nutshell

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$\mathcal{FL} \neq$ description logics

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- Burden of terminology content maintenance and quality assurance

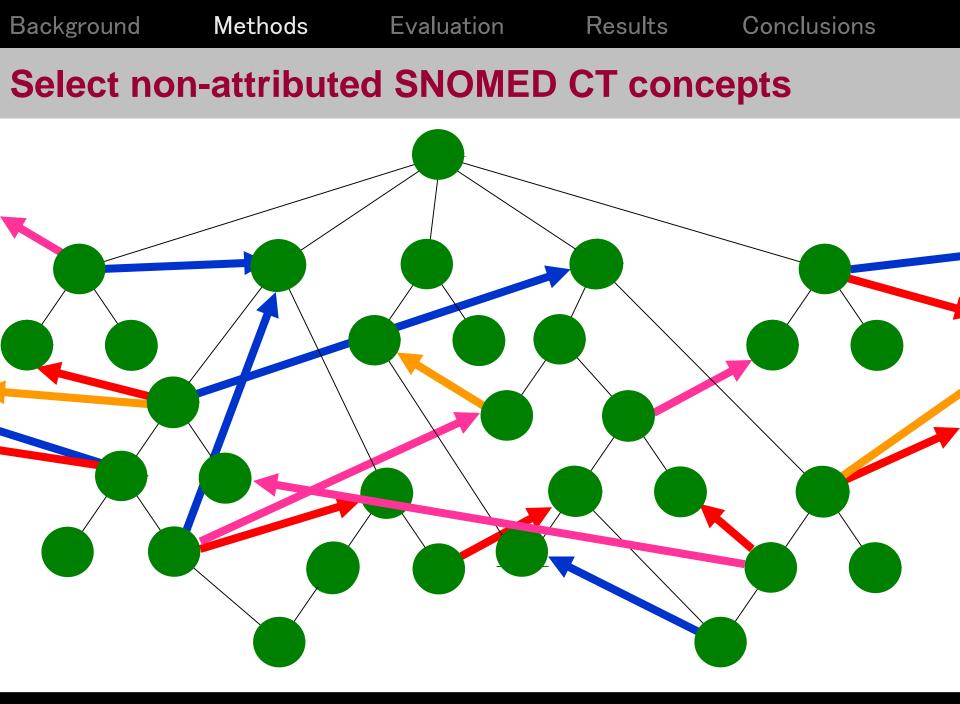
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- To be supported by automated approaches

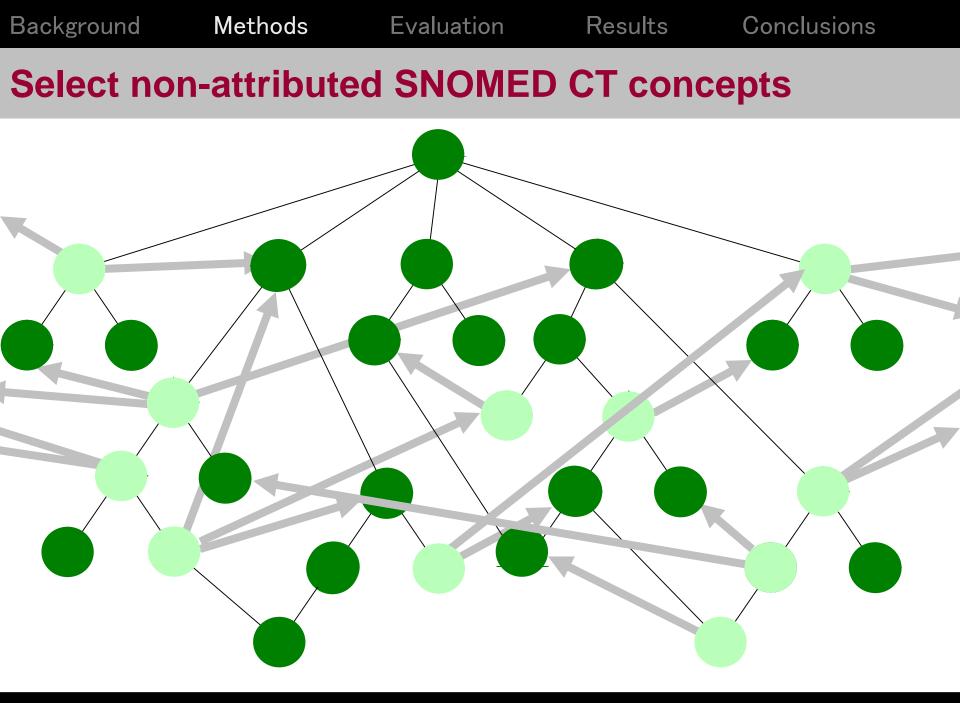
Looking for underspecifications of cross-linkage

- Nearly half (45.2%) of the SNOMED CT concepts (132,125) have no attributes.
- Textual descriptions suggest composed meanings
- Examples:
 - Cerebral function
 - only related to its parent *Nervous system function*
 - expected relation with Brain structure missing
 - Hepatitis notification
 - only related to its parent Disease notification
 - expected relation with *Inflammatory disease of liver* missing

• Source: 01/2009 release of SNOMED CT

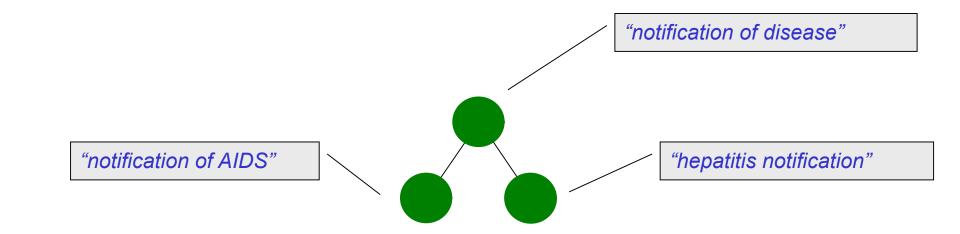
- Source: 01/2009 release of SNOMED CT
- Algorithm:
 - 1. Identify non-attributed concepts





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- Algorithm:
 - 1. Identify non-attributed concepts
 - 2. Extract concept names

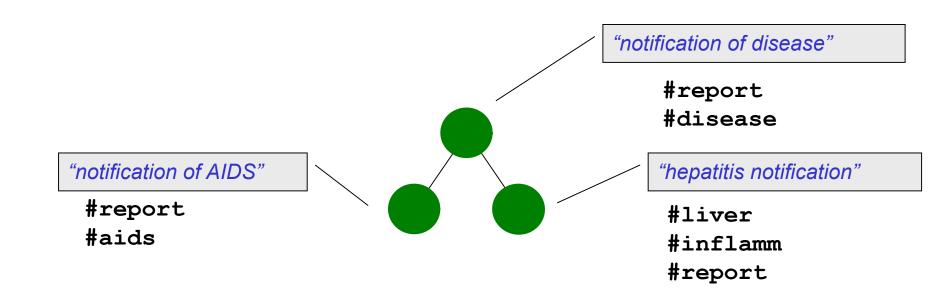




- Source: 01/2009 release of SNOMED CT
- Algorithm:
 - 1. Identify non-attributed concepts
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 - 3. Perform semantic abstraction from word to sets of morphosemantic identifiers (MIDs)

Perform morphosemantic abstraction

Using MorphoSaurus* morphosemantic indexing



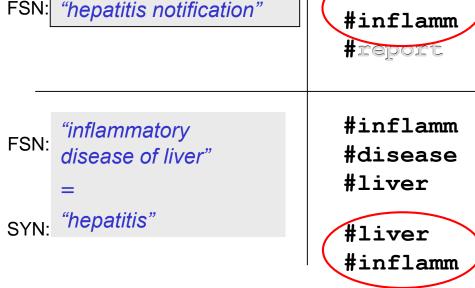
*Markó K, Schulz S, Hahn U: MorphoSaurus - Design and Evaluation of an Interlingua-based, Cross-language Document Retrieval Engine for the Medical Domain. Meth Inf Med 4/2005(44): 537-545.

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 - Match reduced child set against MID representations of all SNOMED descriptions

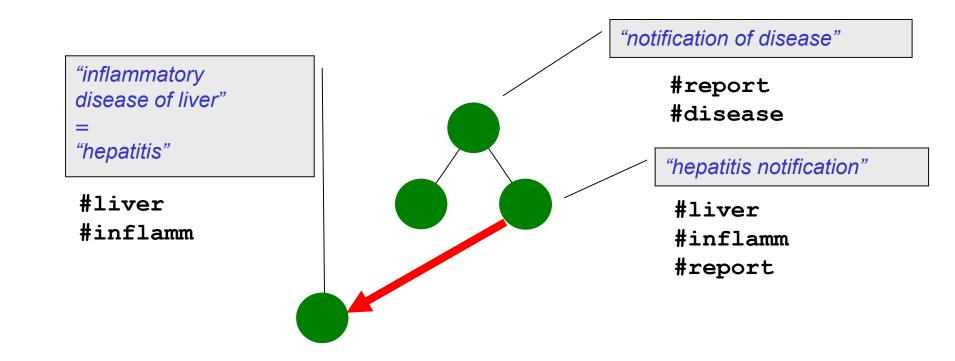
Methods Evaluation Results Conclusions Background Matching heuristics SNOMED MID description For the FSN MID set of every set non-attributed concept: **#report** "notification of disease" FSN: #disease remove MID that occurs in any of this concept's parents check whether the remainder #liver FSN: "hepatitis notification" set coincides with the MID #report representation of some other SNOMED CT concept, *"inflammatory"* considering all descriptions FSN:

- (FSNs, PTs, synonyms)
- consider this concept a refinement candidate



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 - 4. Compare MID sets between children and parents and reduce child sets
 - Match reduced child set against MID representations of all SNOMED descriptions
 - 6. Suggest candidates for refining attributes

Addition of refinement candidate



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Evaluation Methodology

- For each of 14 SNOMED subhierarchies: random sample of 20 underspecified concepts, compared to attribute refinement candidate proposed by the system
- For each of the sample concept verify
 - 1. whether this concept should be refined
 - 2. whether one of the suggested refinement candidates can be plausibly used for refinement.
- Performed by two domain experts. Double rating for interrater agreement measurement: 25%

Results of retrieval experiments

Results of retrieval experiments

	Non-attributed Concepts			Refinement candidates		Analysis of samples(n=20)		Sample based estimation	
	Active	n	%	n	%	refine-	correct	refinable	with
	Concepts					ment	suggest	concepts	correct
SNOMED hierarchies						justified	ion		suggest- ions
Organism	31840	31840	100.0	4973	15.6	0%	0%	0	0
Substance	23554	23554	100.0	8627	36.6	55%	35%	4700	3000
body structure	25637	22386	87.3	15076	58.8	5%	0%	800	0
qualifier value	8823	8823	100.0	3533	40.0	0%	0%	0	0
observable entity	7885	7885	100.0	3647	46.3	70%	50%	2600	1800
Finding	32780	5356	16.3	2253	6.9	90%	75%	2000	1700
physical object	4408	4408	100.0	1339	30.4	85%	80%	1100	1100
morphologic abnormality	4297	4289	99.8	2164	50.4	80%	60%	1700	1300
Occupation	3843	3843	100.0	1330	34.6	75%	10%	1000	100
Product	19310	3541	18.3	686	3.6	100%	60%	700	400
Event	3578	3529	98.6	447	12.5	85%	45%	400	200
Disorder	63874	2812	4.4	1080	1.7	90%	60%	1000	600
Procedure	47764	2256	4.7	1001	2.1	85%	65%	900	700
Others	14511	7603	52.4	2396	16.5	75%	60%	1800	1400
TOTAL	292104	132125	45.2	48552	16.6			18700	12300

BackgroundMethodsEvaluationResultsConclusionsResults

- Interrater agreement (Kohen's kappa):
 - A concept should be refined: 0.55 (low !)
 - There is a proposed refinement candidate: 0.74
- Estimation: approximately 18,000 SNOMED CT concepts can be refined.
- Problematic suggestions:
 - Macaroni for Macaroni maker
 - Canada for Salmonella canada
 - Acyl carnitine for Acylcarnitine hydrolase
 - First for Female first cousin

(already fully defined by the intersection of *First cousin* and *Female cousin*)

Background Methods Evaluation Results Conclusions Conclusions

- Many SNOMED CT concepts are underdefined and can / should be refined
- The proposed methodology was useful to detect underspecifications
- Large difference between SNOMED hierarchies re harvesting and approval of refinement candidates
- "Grey areas"
 - many proposed refinements are debatable
 - only part of refinement candidates not retrieved due to restrictions of the methodology
- Should be considered for future SNOMED CT editing policies

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Stefan Schulz

http://purl.org/steschu

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