Use Cases for Imperfect Knowledge in Health Informatics

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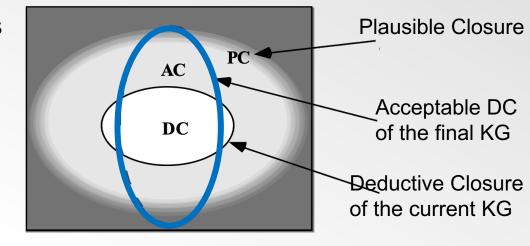
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Use case: Expanding Closure of Medical KG

- Expand Closure of Medical Knowledge Graphs (KG) [1]
 - Plausible inferencing: does not guarantee correctness
 - Probabilistic, abductive, inductive, analogical, ...
 - Expanding deductive closure to acceptable closure
 - Knowledge is partially incomplete & incorrect
- Multi-Strategy Semantic Inferencing [1]
 - Based on Multi-Strategy, Tutor-based Learning (Tecuci [2])
 - Justification Trees are presented as answer to a query
 - Integrate deductive closure with plausible inferences for answer
 - > Explain supporting (plausible) inferences to domain expert
 - ➤ Tutor paradigm: create AC by generalizing trees & validation by tutor



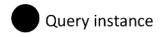


Explanation-based generalization

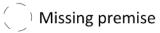
- Step 1: generate deductive justification for queries
 - Implemented via backward chaining

Query: recommend(p01, Azithromycin) Initial Justification Tree: recommend(P, Azithromycin) : — type(P, Adult), type(P, NoLiverDisease), recommend(p01, Azithromycin) illness(P, ChronicBronchitis), illness(P, (a Pneumonia)). illness(P, ChronicBronchitis): - symptom(P, (a Cough)), symptom(P, Fatigue), symptom(P, (a Chills)), deduction symptom(P, SlightFever) . illness(p01, ¢hronicBronchitis) type(p01, Adult) type(p01, NoLiverDiseasePatient) illness(p01, (type Pneumonia)) deduction symptom(p01, (type Cough)) symptom(p01, SlightFever) symptom(p01, Fatigue) symptom(p01, ShakingChills)











Analogy-based Reasoning

- Two entities similar in one aspect, are likely similar in another
 - Transfer knowledge from well-known entity to lesser-known one
- Plausible analogical rules:

```
illness(P, (type Pneumonia)) :~ type(P, ReducedImmunityPatient), symptom(P, (type ShortnessOfBreath)), symptom(P, (type Fever)), symptom(P, ShakingChills)
```

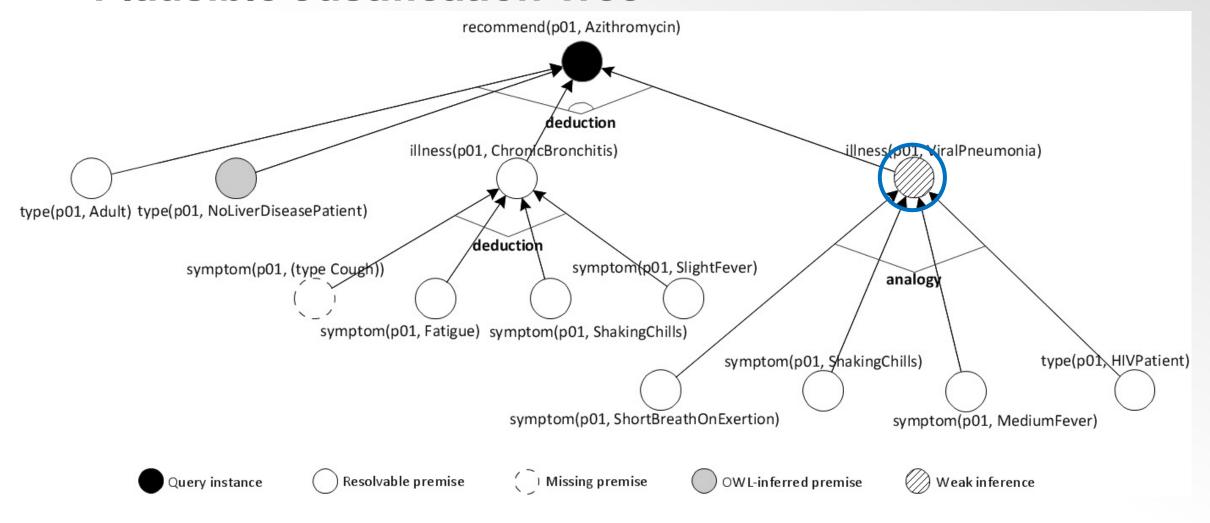
I.e., **if** two patients have *reduced immunity*, share a particular kind of *shortness of breath* & *fever*, and have *shaking chills* **then** they will likely share the same kind of *pneumonia*

```
cds:p07 rdf:type cds:OrganTransplantPatient ; cds:symptom cds:ShortBreathOnExertion;
cds:symptom cds:MediumFever ; cds:symptom cds:ShakingChills ; cds:illness cds:ViralPneumonia
```

```
cds:p01 rdf:type cds:HIVPatient ; cds:symptom cds:ShortBreathOnExertion;
cds:symptom cds:MediumFever ; cds:symptom cds:ShakingChills → cds:symptom cds:ViralPneumonia
```



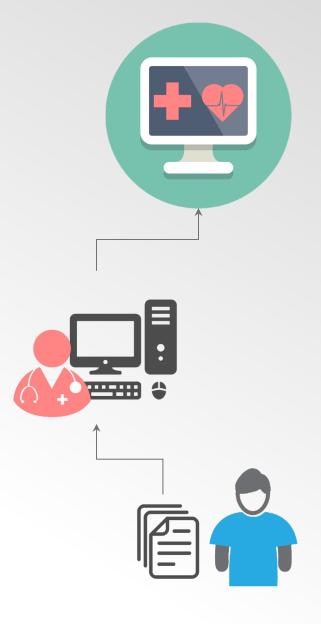
Plausible Justification Tree





Use Case: Plausibly Infer Missing EMR Data

- EMR often show incomplete picture
 - Fragmented and disconnected EMR [3]
 - Different vendors: Telus, QHR, Nightingale, Practimax, Accuro, ...
 - Data input problem from clinician end [3]
 - Hugely increased workloads and burnout using EMR
- Varian Aria Oncology Information System
 - Missing causal associations between diagnoses, treatments
 - It was possible to enter this data in UI but not mandatory (mostly missing)
 - No training data for ML methods
- ➤ Use plausible reasoning over medical KG [4]
 - Leverage medical taxonomies (SNOMED-CT, ICD)
 - Plausibly infer causal relations between diagnoses, treatment





Extract-Transform-Load (ETL) pipeline

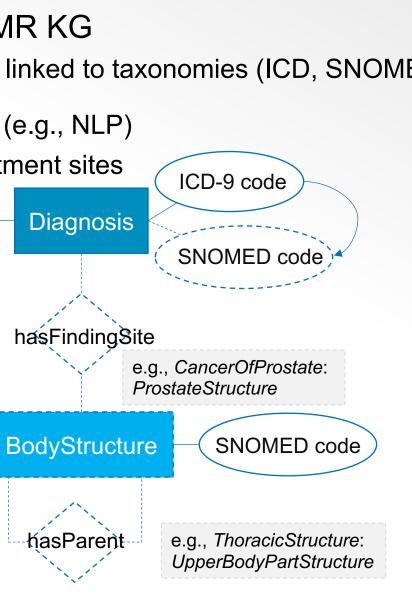
- Transform: EMR data → EMR KG
 - ☐ Oncology EMR: OIS ontology linked to taxonomies (ICD, SNOMED)
- Transform free-text into KG terms (e.g., NLP)
 - ☐ Oncology EMR: free-text treatment sites

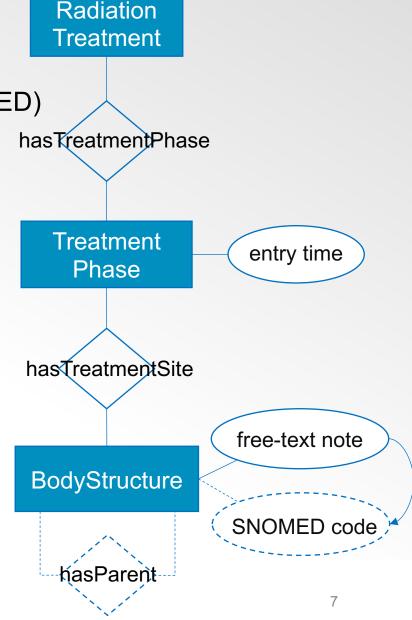
entry time

Enrich with medical KG relations

- Map terminology codes
- Enrich with associational relations
- Enrich with hierarchical relations

SNOMED-CT

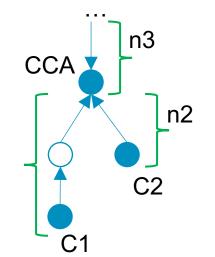


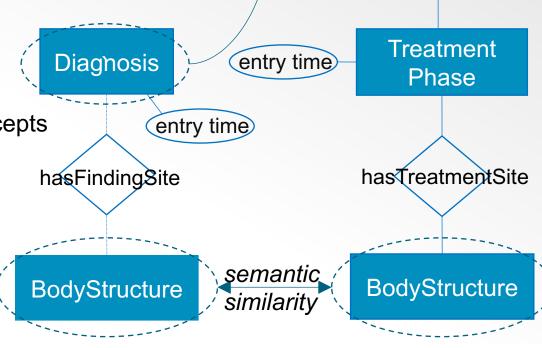


Plausible Metrics for Causal Associations

- Plausible inference pattern based on semantic similarity
 - If some related concepts are similar, then target concepts are causally related
 - 1) Identify related EMR concepts for similarity analysis
 - In EMR KG, concepts have associated body structures
 - 2) Apply semantic similarity on related concepts
 - Wu-Palmer [5]: 2 x n3 / (n1 + n2 + (2 x n3))
 - CD = distance between two concepts and their CCA

Same CD for lower concepts is better than higher concepts





causal

relation

Radiation

Treatment

hasTreatmentPhase

Conclusions

- Two HI use cases for plausible reasoning
 - PhD project: use justification trees to integrate deductive & plausible reasoning
 - Explaining to domain expert why system offers certain answers
 - Industry project: leverage EMR KG & plausible reasoning to solve problem
 - No large-scale dataset available for statistical analysis
- See references for more on these use cases



Demo: Multi-Strategy Inferencing

https://youtu.be/xDj6y0cCjAs



Missing EMR Data: Example Path (1)

- Treatment
 - Site: Structure of vertebral column ("T3-T6 SPINE")
- Diagnosis
 - Type: Secondary malignant neoplasm of brain and spinal cord
 - Site: Spinal cord structure (other: Brain structure)
- Similarity: 0.89
- Path:

Structure of vertebral column and/or spinal cord (depth: 7)

Spinal cord structure

Structure of vertebral column



Missing EMR Data: Example Path (2)

- Treatment
 - Site: Breast anatomy ("RT BREAST")
- Diagnosis
 - **Type**: Malignant neoplasm of lower-outer quadrant of female breast
 - Site: Lower outer quadrant of female breast
- Similarity: 0.86
- Path:

Breast anatomy (depth: 5)



Female mammary gland structure



Lower outer quadrant of female breast



References

- [1] Multi-Strategy Semantic Web Reasoning for Medical Knowledge Bases. Van Woensel, W., Mohammadhassanzadeh, H., Abidi, S.R., Abidi, S.S.R.. In: International Workshop on Biomedical Data Mining, Modeling, and Semantic Integration (BDM2I2015). pp. 1–12. CEUR-WS.org, Bethlehem, Pennsylvania (2015).
- [2] Plausible Justification Trees: A Framework for Deep and Dynamic Integration of Learning Strategies. Tecuci, G. In: Mach. Learn. 11, 237–261 (1993).
- [3] **A Crisis in Healthcare: A Call to Action on Physician Burnout**. A. K. Jha, A. R. Iliff, et al. In: Massachusetts Medical Society, Massachusetts Health and Hospital Association, Harvard T.H. Chan School of Public Health, and Harvard Global Health Institute.
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- [5] **Verbs semantics and lexical selection**. Wu, Z., Palmer, M. In: 32nd annual meeting on Association for Computational Linguistics. pp. 133–138. Association for Computational Linguistics (1994).

