

Automatic Extraction of Skin and Soft Tissue Infection Status from Clinical Notes

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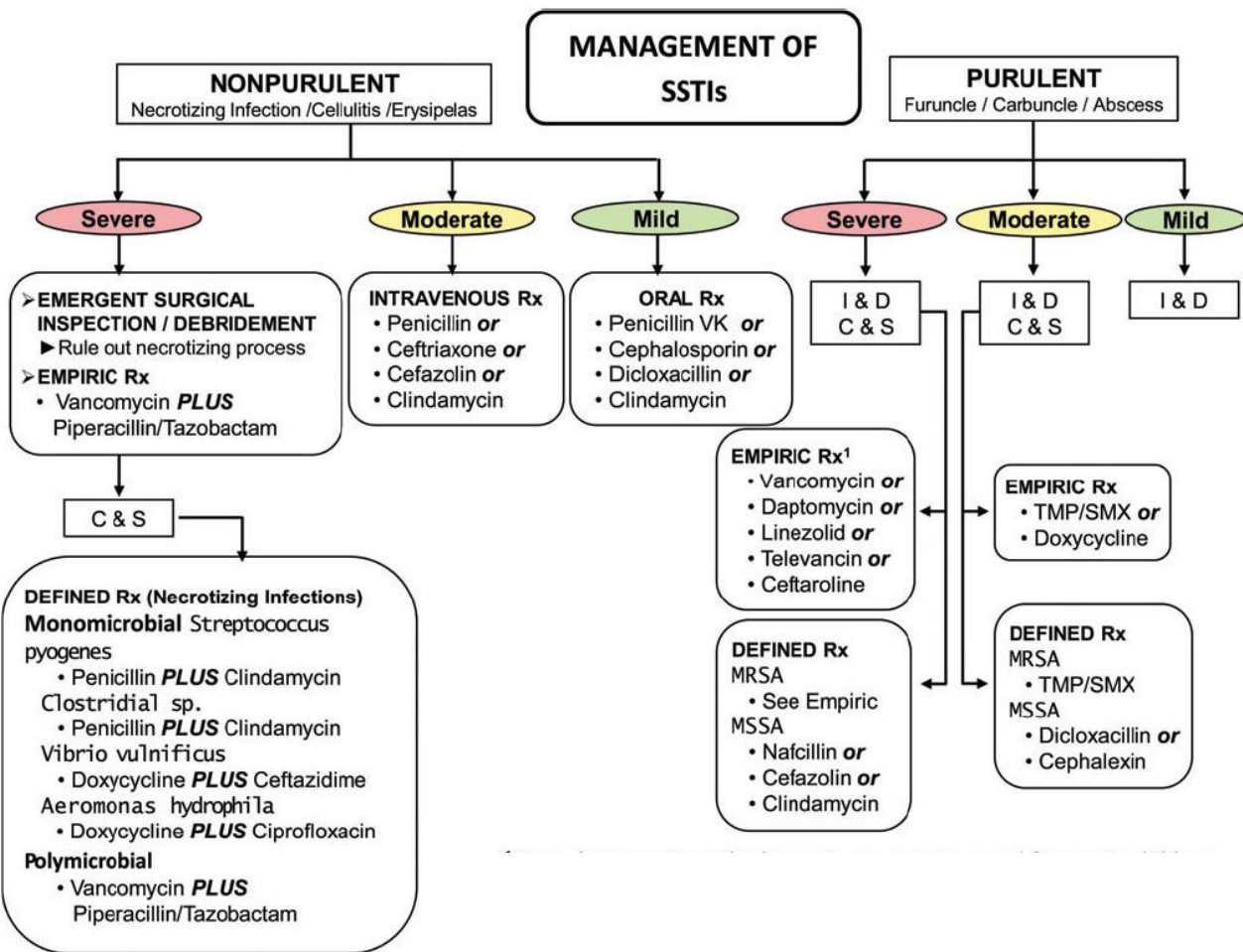
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Skin and Soft Tissue Infections (SSTIs)





Purulent Infections

Infections of the skin and subcutaneous tissues associated with purulent drainage (pus) or drainable fluid (e.g. abscesses, carbuncles)

Non-Purulent Infections

Infections of the skin which are defined as a diffuse, superficial, spreading infection not associated with purulent drainage or fluid collection

IDSA GUIDELINE

Practice Guidelines for the Diagnosis and Management of Skin and Soft Tissue Infections: 2014 Update by the Infectious Diseases Society of America

Why not just use structured data?

- Structured data (ICD-9 and -10 codes insufficiently granular)
- Purulent and non-purulent STIs cannot be captured by ICD codes alone

Purulent Infection

- *There is a fluctuant nodule on the leg*
- *A/P abscess*
- *pt has probable small furuncle of some form on nose*

Non-Purulent Infection

- *A/P probable cellulitis*
- *Suspected cellulitis of left arm*
- *pt has paronychia of the right thumb*

NLP is required to extract purulent vs non-purulent status

Structured Data

<u>ICD -9 or -10 Code</u>	<u>Description</u>
035, A46	Erysipelas
566, 674.1, H00.03, H60.0, J34.0, K61.0, N48.21, N61.1, N76.6, 091.1	Perianal, breast, eyelid, external ear, nose, or genital abscess
367.01, H05.01, H60.1, N48.22	Cellulitis of orbit, external ear, or genitalia
680	Carbuncle and furuncle
681	Cellulitis and abscess of finger or toe
682	Other cellulitis or abscess

Data and annotation

- Cohort consisted of 2,000 randomly selected US Veterans Affairs patients with an SSTI-related ICD code between 2005 and 2018, yielding 6,576 notes:
 - physician notes
 - physician assistant notes
 - nurse practitioner notes
 - podiatry notes
 - dental notes
- Annotation scheme developed:
 - Mention and document level annotations
 - 400 clinical notes (5 batches of 80)
 - Expert annotators (1 board-certified dermatologist; 1 pharmacist; 1 medical student)
 - Annotation guidelines developed (F-score > 0.9)
- Annotated corpus
 - 6,576 notes (3,288 training; 3,288 test)

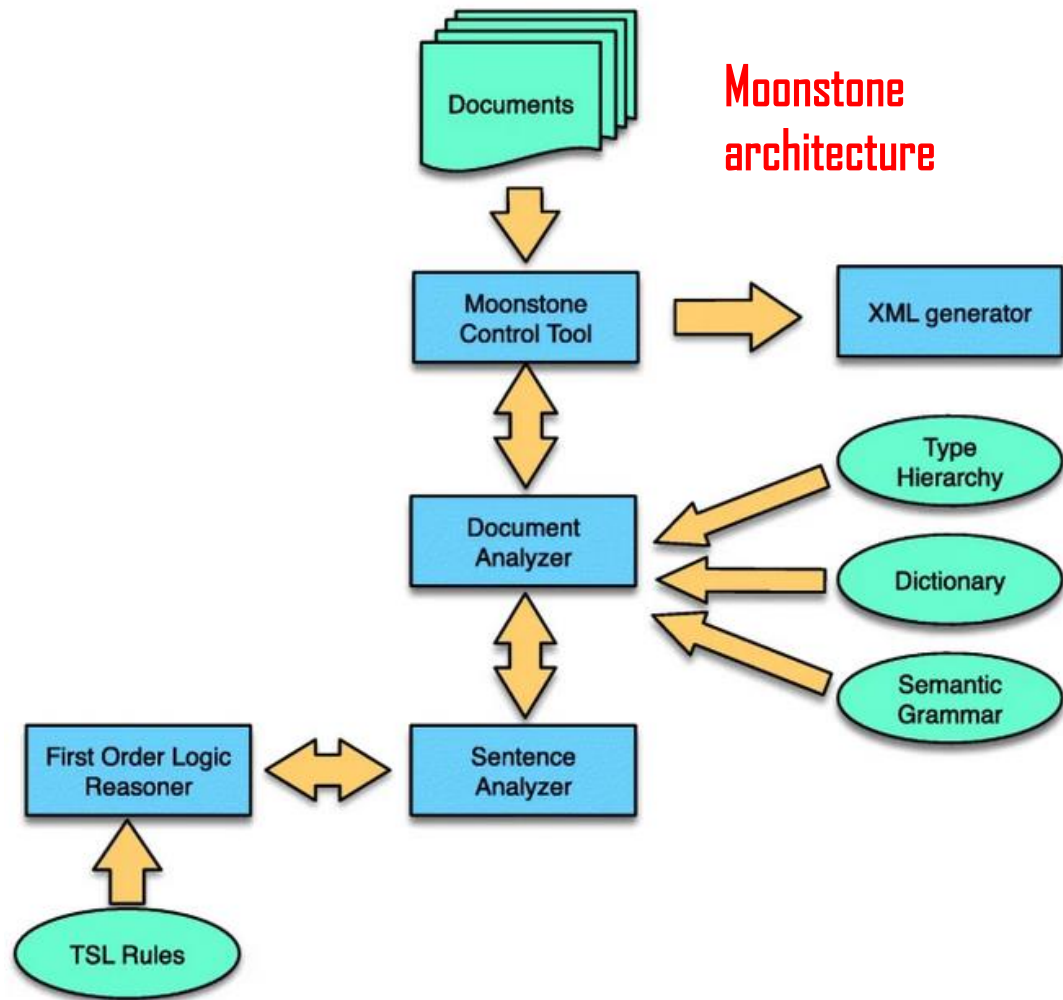


VA

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of Veterans Affairs

Moonstone


- Moonstone rule-based NLP system developed originally for identifying SDoH from clinical notes
- Open-source, Java-based
- System is based around a semantic grammar, a lexical dictionary, a named entity recognition module, and a grammar analysis module
- System advantages: rule-based system allows for rapid development, auditing, and portability
- System disadvantages: requires skill to write rules; complexity of resulting rule set



Results – mention level

- F-score for identifying purulent and non-purulent relatively high
- Performance did not degrade substantially compared to training and validation data

Class	<u>Training</u>			<u>Validation</u>			<u>Testing</u>		
	Prec	Rec	F	Prec	Rec	F	Prec	Rec	F
Purulent	0.77	0.86	0.81	0.71	0.85	0.78	0.7	0.85	0.76
Non-purulent	0.82	0.80	0.81	0.79	0.80	0.80	0.79	0.80	0.80



Results – document level

Attribute		<u>Training</u>			<u>Validation</u>			<u>Testing</u>		
		Pre	Rec	F	Pre	Rec	F	Pre	Rec	F
SSTI	purulent	0.79	0.85	0.82	0.75	0.88	0.81	0.75	0.90	0.82
	non-purulent	0.91	0.74	0.81	0.89	0.71	0.79	0.91	0.77	0.84

- Relatively high F-score for both purulent and non-purulent document-level classification

Other points

- Also extracted *non-specific SSTIs* and *incision and drainage procedures*
- Hardware requirements limited

Limitations & future work

- Compare to neural NLP methods

Contributions:

1. Evidence that an existing rule-based NLP system – Moonstone – can be adapted to a new, unrelated task
2. The development of an NLP tool that can be utilized for clinical, operational, and epidemiological goals related to automated SSTI type detection



Thanks!



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