Combining NLP and Tabular Features

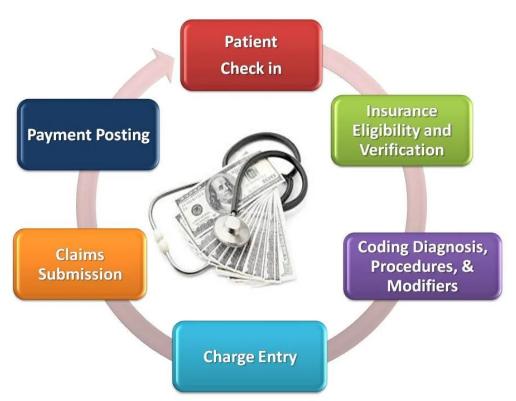
Mapping Clinical Entities to the relevant Section Headers

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About Me

- → Data Scientist III at Episource India Pvt Ltd
- → NLP at large scale on Clinical Text
- → Some recent projects:
 - Clinical entities mapping to relevant section headers
 - Breaking clinical documents into appropriate report boundaries
 - Domain specific ontology search for diseases

The cycle of Medical Coding



Refer to the link for more info -

https://www.m-scribe.com/blog/6-signs-you-have-a-great-medical-billing-process

Objective of the project

- → Section Headers play a critical role in determining the relevance of the Disease
- → Incorrect Disease identification from an irrelevant section can lead to revenue loss of the Insurers
- → Identification alone is not helpful for the project
- → Relevant mapping is pivotal for retention / elimination of Disease for claims
- → Project was built to identify the accurate section headers and map relevant entities with them

Impact of Section Headers

- → Determine if the section is acceptable for capturing code
- → Determine if the condition is current or historic
- → Determine if the clinical document follows a SOAP format for capturing information
- → Incorrect mappings can lead to the following consequences:
 - Additional payments leading to revenue loss Type 1 errors
 - Incorrectly rejected claims Type 2 errors

Initial Approaches

	_Sentence Classification Approach	CRF Approach	Metadata Approach 1. Tabulate feature set 2. Convert categories to integer 3. Feature selection 4. Training and Testing your ML model		
Approach Overview	 Tokenizing the text Converting all the token to lower case Stemming / Lemmatization Stop words removal Vectorization with N-Gram Training and Testing your ML model 	1. Tokenizing the text 2. Converting all the token to lower case 3. Stemming / Lemmatization 4. Stop words removal 5. Vectorization with N-Gram 6. Training and Testing your CRF model			
Pros	Using TFIDF rare words gain more importa Easy on computation	1. Gives importance to sequence in the contex 2. Not sensitive to data imbalance	Works well with limited set of data for the given use case No need for complex DL models		
Cons	Its a BOW approach Ignores semantics and sequence	High computational complexity Does not work well with unknown words	I. Ignores text content totally Overlap of features in both categories		
Best Use Case	Identifying most important keywords	Named Entity Recognition	Any tabular / structured data		
		Shortlisted Approach			

Drawback of the existing approach

- → Section Headers are not sequence problems
 - Chief Complaint Relevant Section Header
 - ◆ Patient is taking medications regularly for all the conditions listed in chief complaint section
- → Too many Type 1 errors
- → Text sequence may not match the PDF format
 - Section headers are usually not a part of any sentence

Examples

Additional Risk Factors:

Fall Risk

Within the last 12 months, have you fallen or experienced any difficulties with balance or walking? Yes
Abuse & Neglect

Do you feel safe at home? Yes

In the last year, have you had unwanted touching, felt physically threatened or abused physically, sexually, emotionally, financially, or psychologically by a partner, spouse, family member, or another? **No**

Suicide Risk

Have you had thoughts of harming yourself or others? No

Anxiety

Do you find that you can't stop worrying'? Yes

Within the past 2 weeks have you consistently felt restless or on edge? Yes

Depression

In the past 2 weeks:

Have you felt down, depressed or hopeless about your life? No

Have you lost interest in doing things you used to enjoy? No

Review of Systems

See HPI All other systems are negative on ROS.

Section Headers incorrectly classified as Disease

> Incorrect Header

Additional Challenges

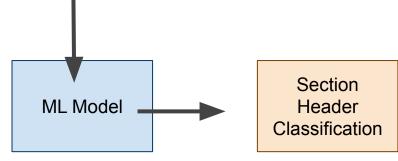
- → Data from a table within PDF chart
- → Table headers
- → OCR errors during PDF to Text conversion
- → Alignment and text sequencing mis-match

The Solution

- → Combine Text and Metadata Features:
- → Text features:
 - Lower case text
 - Match with known headers
 - Spell correction
- → Metadata:
 - Font size
 - Font style
 - Position info

Sample of final matrix passed to ML Model

Sentence 1	Word 1	Word 2	Word 3	Metadata 1	Metadata 2	Metadata 3
Sentence 2	Word 1	Word 2	Word 3	Metadata 1	Metadata 2	Metadata 3
Sentence 3	Word 1	Word 2	Word 3	Metadata 1	Metadata 2	Metadata 3



Mapping Example

Impression & Recommendations:

Problem #1: Physical exam, routine (ICD-V70.0) (ICD10-Z00.00)

Reviewed preventive care protocols, scheduled due services, and updated immunizations.

Orders:

Comp Metabolic (CMP)
Lipid Panel (LPDPA)
CBC No differential (CBCND)
Thyroid Cascade (TSHRC)
Hemoglobin A1C (HA1CG)

Problem #2: Tics of organic origin (ICD-333.3) (ICD10-G25.69)

Will refer to neurology as pt has neurosurgeon but not neurologist. Discussed with pt having consult for the tics as been present for years. Pt stated didnt improve with shunt. Pt does not feel any medication has worsened it.

Orders:

Neurology Consult & Treat (*)

Problem # 3: Obesity (ICD-278.00) (ICD10-E66.9)

Encouraged cont diet and exercise.

Will start on wellbutrin to see if helpful with weight loss and pt symptoms of what she feels is ADHD. Discussed returning to weight loss clinic for medication management as she is fearful of sleeve.

Mapping Constraints

- → Identification
- → Position
- → Isolation
- → Page Layout