LOINC Document OntologyPipeline Documentation

OVERVIEW

- Metadata Pre-processing
 - Complexity reduction from Millions to Thousands
- Bag of Words Implementation
 - Mapping Bag of words to LOINC Codes
 - Mapping LOINC Codes to Unique Bag of words
 - Mapping LOINC Codes to individual notes
- Results & Analysis + Discussion

Num of	Number of	
Parts	Notes _	% ■
0	7,225,678	5%
1	25,015,785	19%
2	41,216,686	31%
3	33,005,516	25%
4	21,267,870	16%
5	4,899,873	4%
Total	132,631,408	

A. Metadata Preprocessing

A.1 Notes Metadata Consolidation

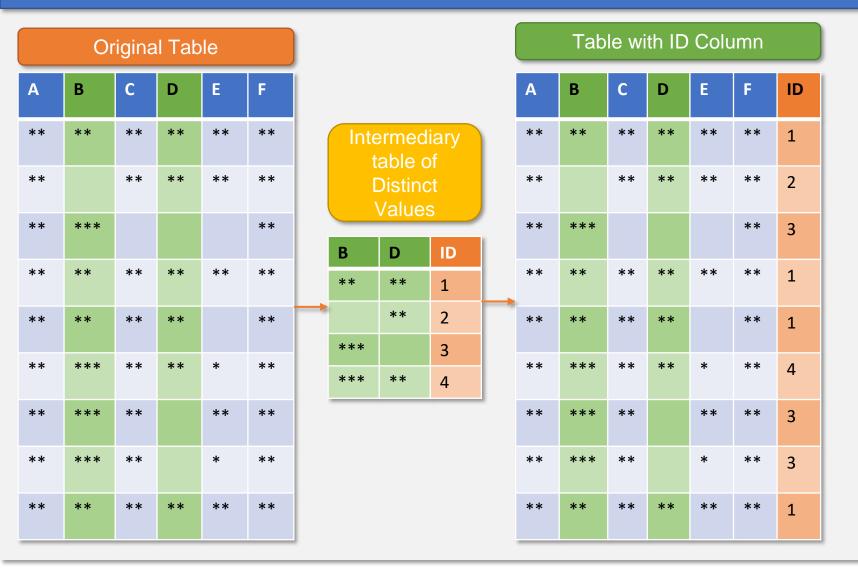
Complete Table of all Decompressed notes (132.6 Million Notes Metadata)

Distinct Values for all columns containing relevant data (29M)

Remove irrelevant words from columns to decrease unique combinations EX. Doctor A is prescribing a medicine Doctor B is prescribing a medicine Doctor C is prescribing a medicine Doctor, prescribing, medicine **UMLS CUI Terms** After removal of irrelevant words, the unique number of columns becomes around 289,000

- Our notes number in around 132.6 Million.
- We can get distinct value of important columns, which decreases number to around 29M.
- Next, we can remove irrelevant words (words that don't help us connect note to LOINC Codes).
- Final number of unique notes comes out to be around 289,000 which is iterable in under 2 hours.

A.2 Data Table Preparation



- The ID field will act as the primary key, allowing for efficient joining of tables with correlated LOINC Codes.
- We will operate on the significantly smaller Intermediary Table and join back to the individual notes at the end.

A.3 Bag of Words table with ID



- Using the intermediary table from last step, we will create a new Table.
- The table will have a Bag
 of Words field, storing all
 the unique bags of words.
- An ID field will carry over.
- A new **LOINC Code** field, to be programmatically populated.

B. Bag of Words Implementation

B.1 Vectorized Implementation

{'research': 0, 'exam.hypertension': 0, 'midwife': 0, 'pulmonary': 0, 'rectal': 0, 'hernia': 0, 'student': 0, 'lo 'thromboembolism': 0, 'program': 0, 'exam.feet': 0, 'sleep': 0, 'exam.extremity': 0, 'intestines': 0, 'multi-spec 'intervention': 0, 'microbiology': 0, 'encounters': 0, 'nutrition': 0, 'exam.genitourinary': 0, 'smoking': 0, 'su 'exam.hiv-related': 0, 'exam.sense': 0, 'rehabilitation': 0, 'specialist': 0, 'telephone': 0, 'encounter': 0, 'an 'hepatology': 0, 'exposure': 0, 'neurological': 0, 'breast': 0, 'obstetrics': 0, 'child': 0, 'tumor': 0, 'confere 'comprehensive': 0, 'exam.cushings': 0, 'autopsy': 0, 'labor': 0, 'exam.multiple': 0, 'surgical': 0, 'preoperativ 0, 'equipment': 0, 'reproductive': 0, 'hospital': 1, 'cardiovascular': 0, 'pharmacogenomics': 0, 'cardiopulmonary 'breastfeeding': 0, 'otolaryngology': 0, 'community': 0, 'candidate': 0, 'mellitus': 0, 'term': 0, 'restrictive': .audio': 0, 'interdisciplinary': 0, 'gulf': 0, 'exam.skin': 0, 'vascular': 0, 'toxicology': 0, 'gaps': 0, 'perina 0, 'pharynx': 0, 'fingers': 0, 'licensed': 0, 'family': 0, 'exam.general': 0, 'colon': 0, 'aerodigestive': 0, 'an 'primary': 0, 'trauma': 0, 'exam.aid': 0, 'gall': 0, 'therapeutic': 0, 'military': 0, 'assessment': 0, 'digestive 'adhesions': 0, 'shift': 0, 'reconciliation': 0, 'exam.chronic': 0, 'registered': 0, 'nerves': 0, 'exam.mouth': 0 .obstructive': 0, 'gastroenterology': 0, 'environmental': 0, 'individual': 0, 'immunology': 0, 'exam.eye': 0, 'sy 'initial': 0, 'gynecology': 0, 'functional': 0, 'counseling': 0, 'dialysis': 0, 'bladder': 0, 'narcolepsy': 0, 'a 'recovery': 0, 'product': 0, 'orthotics': 0, 'pharmacy': 0, 'parathyroid': 0, 'cleft': 0, 'circulatory': 0, 'diso

	LoincNum	PartNumb	PartTypeName	PartName
38	100446-4	LP173051-6	Document.Setting	Outpatient
39	100446-4	LP420041-8	Document.SubjectMat	Breastfeeding
40	100447-2	LP173418-7	Document.Kind	Note
41	100447-2	LP173213-2	Document.TypeOfService	Progress
42	100447-2	LP173051-6	Document.Setting	Outpatient
43	100447-2	LP268363-1	Document.SubjectMat	Burn management
44	100448-0	LP173418-7	Document.Kind	Note
45	100448-0	LP173213-2	Document.TypeOfService	Progress
46	100448-0	LP173051-6	Document.Setting	Outpatient
47	100448-0	LP207300-7	Document.SubjectMat	Cardiac surgery
48	100449-8	LP173418-7	Document.Kind	Note
49	100449-8	LP173213-2	Document.TypeOfService	Progress

[0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
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0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
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0]																									-1	

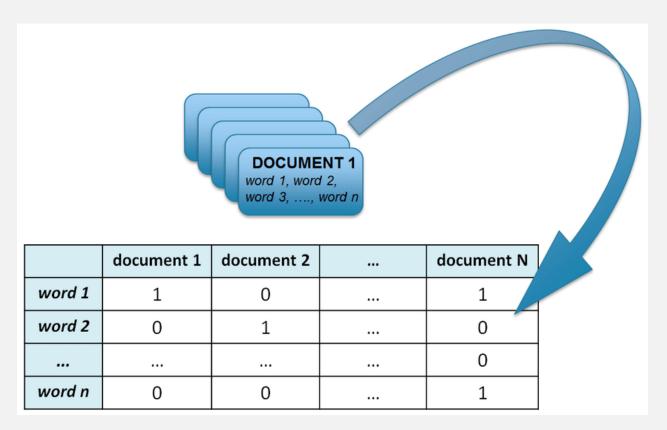
- Bag of words are implemented using Vectors
- Allows for compact storage
- Allows for efficient comparison of two vectors (i.e. Bag of Words comparison)
- Vector size is constant and so doesn't depend on word count.

B.2 LOINC Code Word-Base

	LoincNum	PartNumb	PartTypeName	PartName
38	100446-4	LP173051-6	Document.Setting	Outpatient
39	100446-4	LP420041-8	Document.SubjectMat	Breastfeeding
40	100447-2	LP173418-7	Document.Kind	Note
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- Vector size is constant and dependent on the LOINC Codes.
- DocumentOntology.csv provides a list of LOINC Codes along with part numbers/names
- Using the Part names, we can create a Word-base of unique keywords (burn, note, cardiac etc.)
- The keyword count comes to 519, which is our vector dimension.

B.3 LOINC + Document vectors

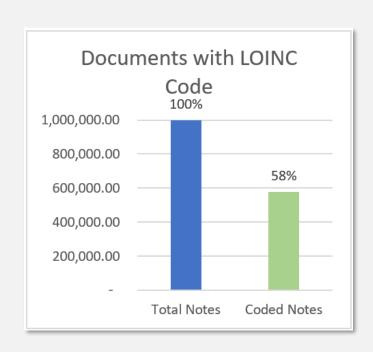


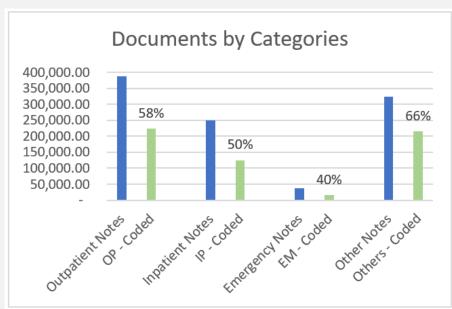
Baj-Rogowska, Anna. (2017). Agile Commerce in the light of Text Mining

- Having defined our vector, we can create vectors for LOINC Codes
- For documents, we will only accept words in LOINC Word-base
- Takes under a minute to convert 1M metadata rows into vectors
- Compare documents to LOINC code using vectors and dot-products.
- Roughly 5-20 times faster than string comparison

C. Results & Analysis + Discussion

C.1 Results and Accuracy





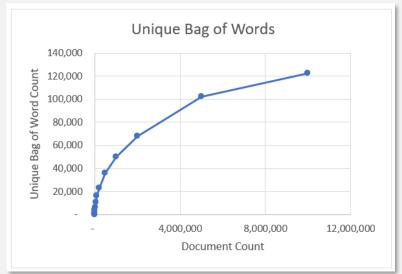
- I tested the pipeline with
 1M Documents.
- Around 58% were mapped to LOINC Codes
- Broken down into categories, the numbers range from 40-66% coverage.
- For many OP/IP
 documents, they are only
 missing Kind of
 Document (KOD) (i.e.
 Note, Letter, summary
 etc.)

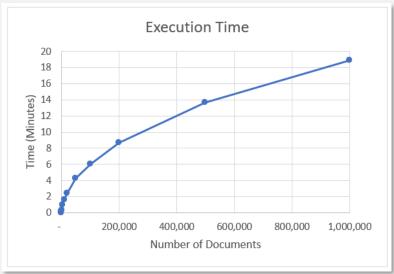
C.2 Values for Categorization

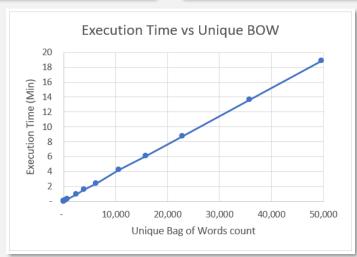
Outpatient Types	Inpatient Types	Others -	Emergency
UH OUTPATIENT	UH INPATIENT	Clinic	UH EMERGENCY
WCH PHYSICIAN OP CLINIC	Inpatient	Between Visit	WCH EMERGENCY PATIENT
Outpatient	WCH INPATIENT	Historical	Emergency
MOI OUTPATIENT	MISSOURI PSYCH INPATIENT	UH NO TECHBILL	UH PSYCH ER ASSESSMENT
UH AMBULATORY SURG	WCH NEW BORN INPATIENT	UH TRAUMA	LRA ER
EF OUTPATIENT	MOI INPATIENT	UH OBSERVATION	
WCH OUTPT ANCILLARY	UH INPATIENT NBN	UH DIAGNOSTIC TESTING	
WCH SHORT STAY	WCH REHAB UNIT PATIENT	UH DIAGNOSTIC TEST	
MOI AMBULATORY PROCEDURE PATIENT	zzEF INPATIENT	Virtual Care	
Outpatient in a Bed	zzCR GERIATRIC PSYCH IP	Day Surgery	
EF SHORT STAY	LRA IN	UH THERAPY SERIES	
Non-Admit		Recurring	
MOI SHORT STAY		WCH DIAGNOSTIC TESTING	
MISSOURI PSYCH OUTPATIENT		PreClinic	
UH OUTPT CHRONIC CARE MGMT		Observation	

- Sample list of categories entries taken from unique list in ENCNTR_TYPE_CD
- Ranked based on decreasing frequency
- Others may include both inpatient and outpatient documents, but we can't tell based on the info we have

C.3 Complexity





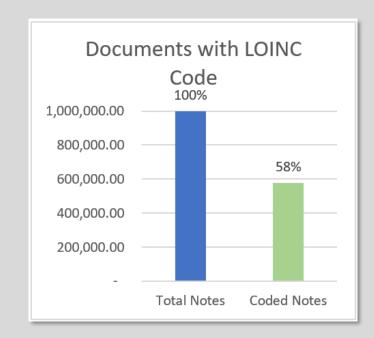


- The algorithm scales linearly with the number of unique Bag of Words, which is a fraction of the number of documents.
- The graph on the left shows the unique bag of word count as a function of document count.
- The graph on the right shows the program runtime as a function of document count.

C.4 Final Thoughts

Notes

- We might be able to significantly boost the coverage by defaulting to 'notes' for documents that don't have a KOD keyword.
- There are LOINC Codes that are 'Discouraged' or 'Deprecated'
 - Very few but identifying and removing them may lower our numbers by a percentage or two



D. END