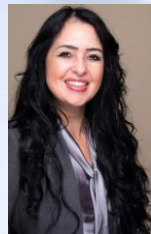


# Transforming the Clinical Decision Support and Clinical Analytics Landscape Through NLP



**Elia Lima-Walton, MD**

Director Data Science & Clinical Analytics Health  
Elsevier

# Agenda

Our mission, how we operate, our community

Elsevier's Medical Taxonomy Introduction & Connection with Health Products

Data Science, NLP, and Clinical Analytics Capabilities

# Our Mission

- Elsevier helps researchers, educators, and healthcare professionals advance science and improve health outcomes for the benefit of society.
- We do this by facilitating insights and critical decision-making for customers across the global research and health ecosystems.



## Who we are

A global information analytics  
business specializing in science  
and health

## What we do it

We help you solve  
your challenges, for  
the benefit of  
humanity



ELSEVIER


## What we do

We help institutions  
and professionals  
progress science,  
advance healthcare  
and improve  
outcomes

## A unique combination

Combine content with technology,  
supported by operational efficiency,  
to turn information into actionable  
knowledge

# Supporting Our Five Main Customer Segments

 <b>Clinicians</b>	'Consider this treatment for this patient'
 <b>Researchers</b>	'This article answers your questions'
 <b>Governments</b>	'This is the research to invest in'
 <b>Pharmaceutical companies</b>	'This is the novel treatment you should pursue'
 <b>Medical and Nursing Students</b>	'This is the area you need to improve to succeed'



# Elsevier Flagship Products

At Elsevier we facilitate insights and critical decision-making for our global customers.

Sherpath<sup>®</sup>

ClinicalKey<sup>®</sup> Student

ClinicalKey<sup>®</sup>

bepress<sup>™</sup>

ScienceDirect<sup>®</sup>

PharmaPendium<sup>®</sup>

ExpertPath<sup>™</sup>



Embase<sup>®</sup>

Pure



SciVal

Science-Metrix

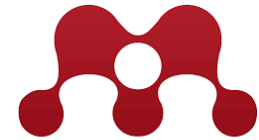
3D4MEDICAL

Reaxys<sup>®</sup>



SciBite  
an ELSEVIER company

STATdx<sup>®</sup>



MENDELEY

Arezzo

Scopus<sup>®</sup>



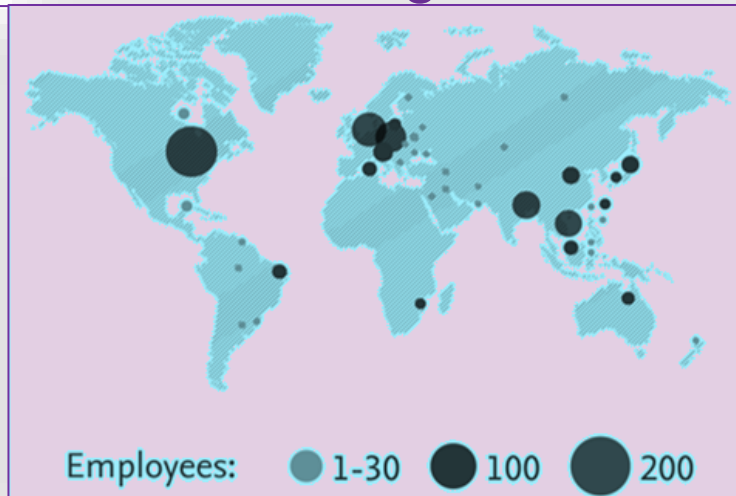
authess

ExpertPath<sup>™</sup>

NLP  
SUMMIT  
SSRN

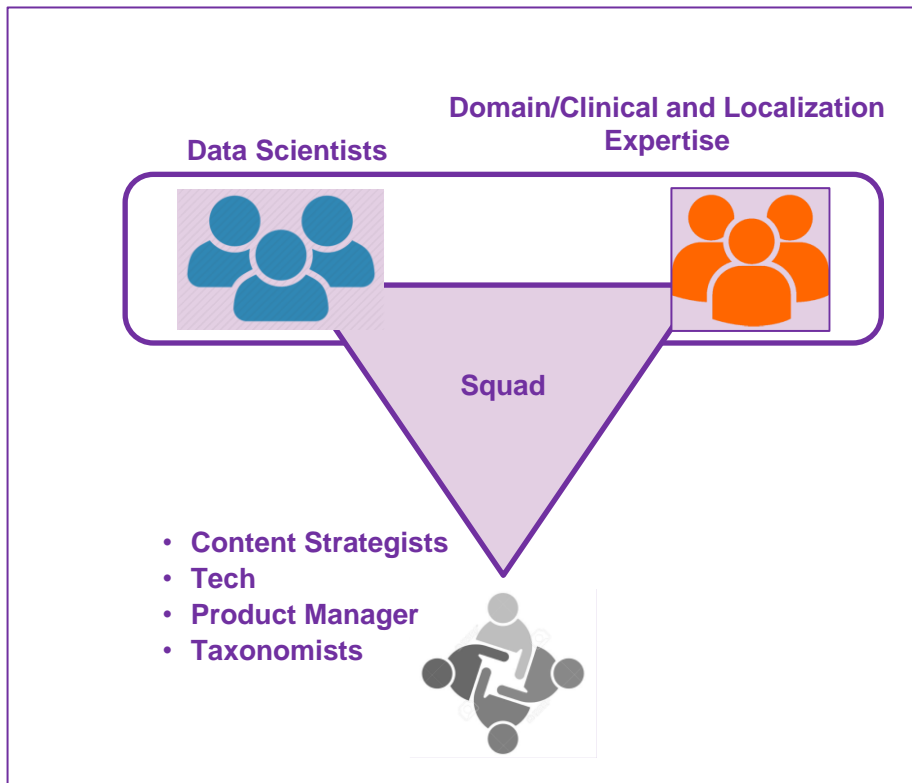
# Elsevier's Data Science Community and Global Organization

- **250+ data scientists and clinician specialists**
- Several Global Collaborations with Universities in the United States, Europe, and India
- Main hubs in:
  - Amsterdam
  - London
  - Philadelphia
  - Chennai
  - Bengaluru
- Artificial Intelligence Lab in Amsterdam  
<https://discoverylab.ai/>



Elsevier has **8,700** employees and serves customers in **180+ countries**

# Data Science Health, Who We Are and How We Operate



- We are organized locally and globally
- Support Health initiatives globally with data science, clinical, and localization expertise
- Manage extended bench of external contractors and suppliers
- Execute on innovative projects aligning with internal external collaborators

- Philadelphia
- Houston
- Dallas
- Knoxville
- Asheville

- Amsterdam
- Barcelona

- Chennai
- Delhi/Gurgaon
- Bengaluru
- Mumbai



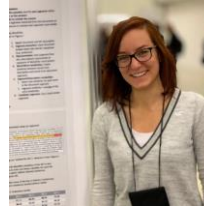
# Meet the team– Data Science & Clinical Analytics



**Elia Lima-Walton, MD**  
Director Data Science &  
Clinical Analytics Health



**Hina Nazir, MD**  
Clinical Analytics Manager



**Dasha Herrmannova, PhD**  
NLP & Data Science  
Manager



**Leah Livingston, RN**  
Clinical Data Scientist &  
Analytics Specialist



**Sameer Chivukula, MSc**  
Data Science Manager



**Sharvari Jadhav, MSc**  
Data Scientist



**Harsh Sindhwa, MSc**  
Data Scientist



**Charlotte Tesselaar, MSc**  
Sr. Localization Specialist



**Pranita Mahajan, MEng.**  
Senior Data Scientist



**Vidhyaa Rajee, MSc**  
Data Scientist



**Dwipjayeta Choudhury, RN**  
Sr. Clinical Data Analyst



**Maria Jedrzejowska, MSc**  
Localization Specialist

**Roles and expertise**  
**Multi-disciplinary department:** Data Scientists, Clinical Analytics Specialists, Localization Specialist, Knowledge Management, Life Sciences  
**Data Science and Health expertise:** Machine Learning, Natural Language Processing, Machine Translation, Optical recognition, Clinical Analytics, Medicine, Nursing, Hospital Admin, Pharmacy and Drug Development



# Key Capabilities Developed in Data Science and Clinical Analytics



# Combining the Power of our Trusted Medical Content with Data Science

## Key capabilities which we prioritize and develop:

### 1.) Search and Discoverability, NLP & ML

***Concept/taxonomy alignment and information retrieval:*** Aligning medical taxonomy and ontology to content in order to power user search experience

***2.) Standardization of metadata, normalization, new technologies:*** Developing innovative technologies to support integration, normalization, and generation of text, images, and multimedia assets.

***3.) Globalization/Localization :Focus on the global customer*** by adapting products, content, and initiatives to various locales and cultures.

# Unique Assets and Capabilities to Extract Information from Clinical Records and Medical Curricula

## Use Cases

- Deidentify text records
- Commercialize health data
- Risks, predictive analytics
- Coding and reimbursement
- Extract knowledge from literature
- Map insights to patients
- Mapping educational curricula

## Problems to Solve

Convert unstructured into structured data from

- clinical notes
- radiology reports
- pathology reports
- lab reports
- biomarkers
- education alignment

Standardize data from multiple hospitals (SNOMED, ICD 10, CPT, LOINC, ATC, NDC, ...)

Contextualize and interpret data

## Solutions

NLP

taxonomies

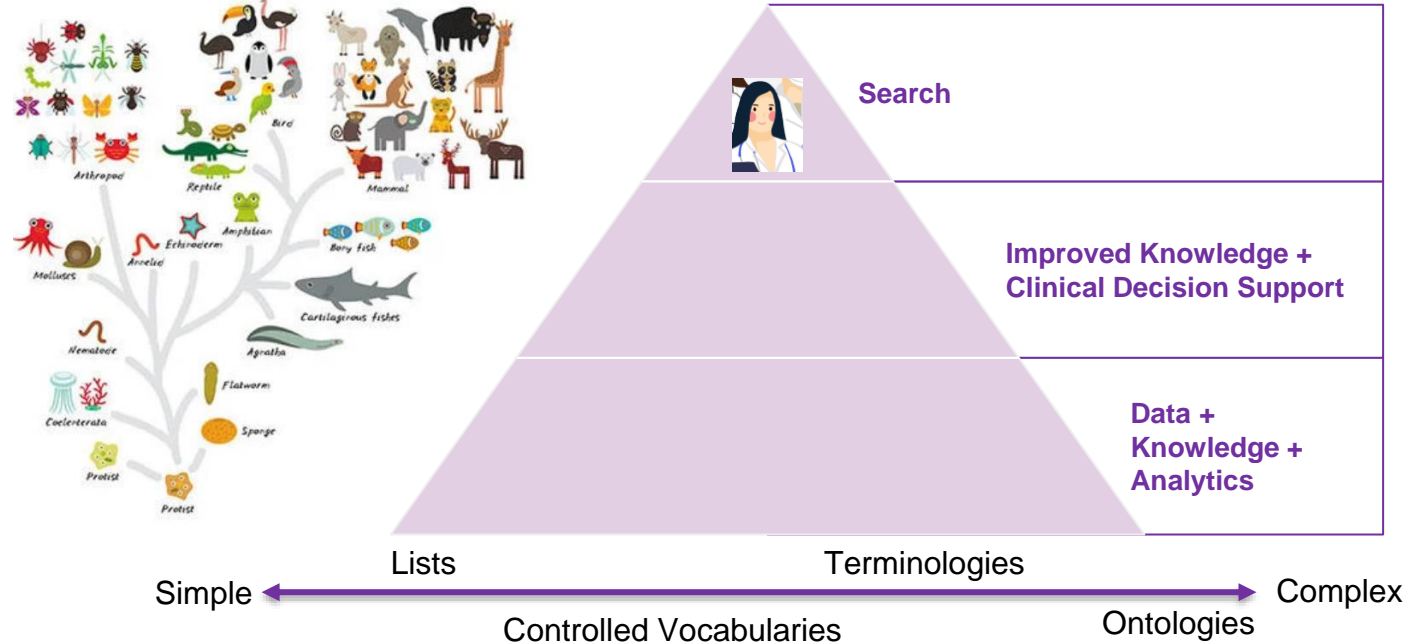
knowledge graph

# Taxonomy for synoptic content discoverability



# Taxonomy Powering Search and Product

Taxonomy: Science of identifying and naming species and arranging them into classification



## Objective

Clinicians, researchers, students, and educators need concise actionable answers. Identify and prioritize key concepts embedded in content to aid in improved information retrieval.

# Introducing Elsevier's Medical Taxonomy

- **Multilingual**: English, French, Spanish, German, Brazilian Portuguese
  - Arabic, Chinese, Somali, Nepali, Armenian, in pilot
- **Concept-based**: All terms, synonyms, translations, mappings are related to one unique identifier ("IMUI")
- **Ontology**: Provides semantic relationships between concepts (symptoms of a disease, treatment procedures of a disease, complications of a disease etc...)
- Explicitly mapped to **international medical standards** (*SNOMED-CT, ICD-10-CM, LOINC, CVX, etc.*) and Elsevier's internal vocabularies (Gold Standard, Dorland's).
- Continuously curated by team of clinical medical terminology experts, available via API
- Is at the heart of our knowledge graph, connecting concepts and relationships supported by evidence in content, unlocking knowledge through scalable, easily-navigable information services.
- Knowledge graph has **400K+** medical concepts, **900K SYN** with over **220K semantic relations, 90K symptoms, 46K drugs, 75K diseases, 63K procedures** and extensive mappings to industry vocabularies and multi-language support.

## Introducing Elsevier's Taxonomy and Knowledge Graph

## Parents

- Breast disorders (IMUI:2805667)
- Cancer of the thorax (IMUI: 8258010)
- Neoplasm of breast (IMUI: 8258030)
- Malignant neoplasm (IMUI: 2791003)

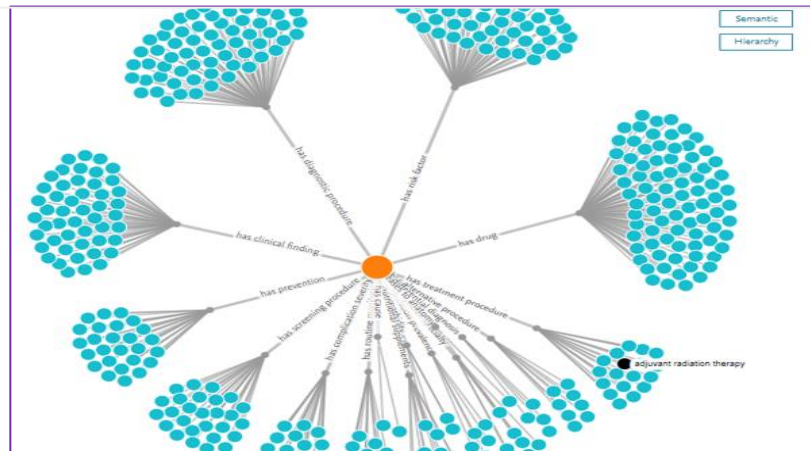
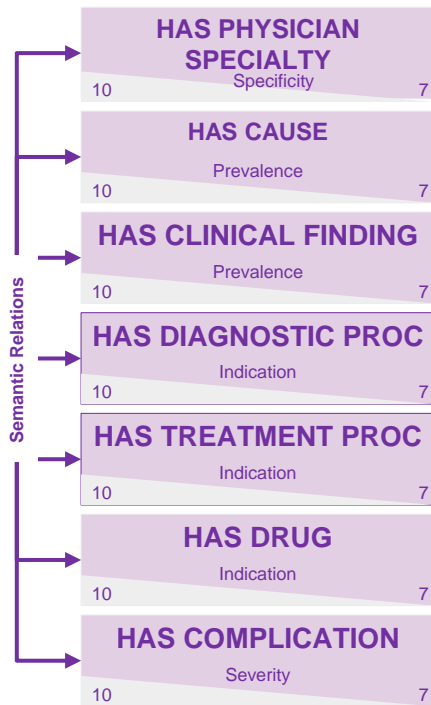
## IMUI: 2790981

- Lang: En
- **Medical Name**
- Malignant neoplasm of the breast
- **Consumer-Friendly Name**
- Breast cancer
- **Synonyms**
- Malignant tumor of the breast
- Malignant breast neoplasm
- Breast CA
- **STY/STYGROUP**
- Neoplasm process / Disease
- **Mappings**
- UMLS – C0006142 (exact)
- ICD9-CM – 174.9 (exact)
- ICD10-CM – C50 (exact)
- SNOMED-CT – 254837009 (exact)
- MeSH – D001943 (exact)

## Children

- Breast sarcoma (IMUI:8258036)
- HER2-positive breast cancer (IMUI: 8316103)
- Male breast cancer (IMUI: 8009640)
- Lobular carcinoma (IMUI: 3816091)

100

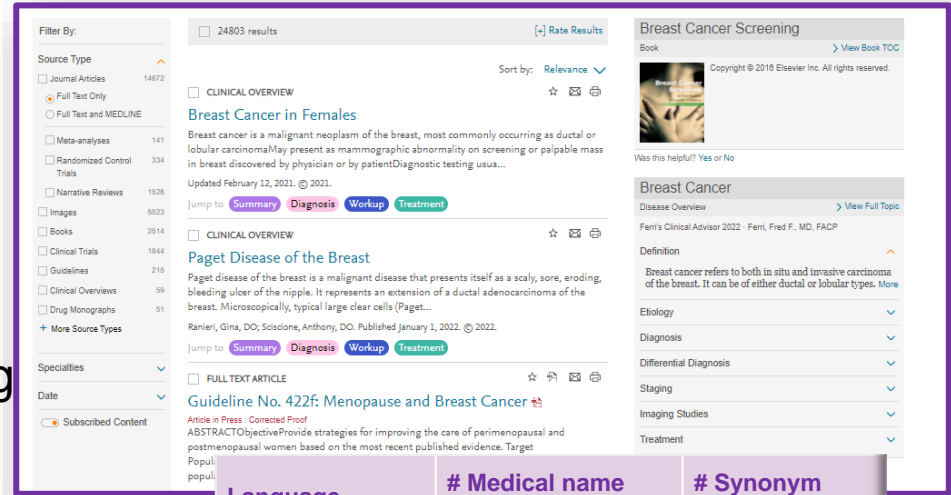


- What clinical findings are associated with **breast cancer**?
- How do you screen **breast cancer**?
- What is the diagnostic work-up for **breast cancer**?
- What is the gold standard treatment for **breast cancer**?
- What are common differential diagnoses for **breast cancer**?

# Helping Clinicians and Students Worldwide

- ClinicalKey US
- ~20 million content assets
- **Our taxonomy is used for:**
- Autocomplete search
- More accurate search results
- Faceted navigation (Related screening drugs)
- Query refinement

Localized versions for France, Spain, Brazil and Germany

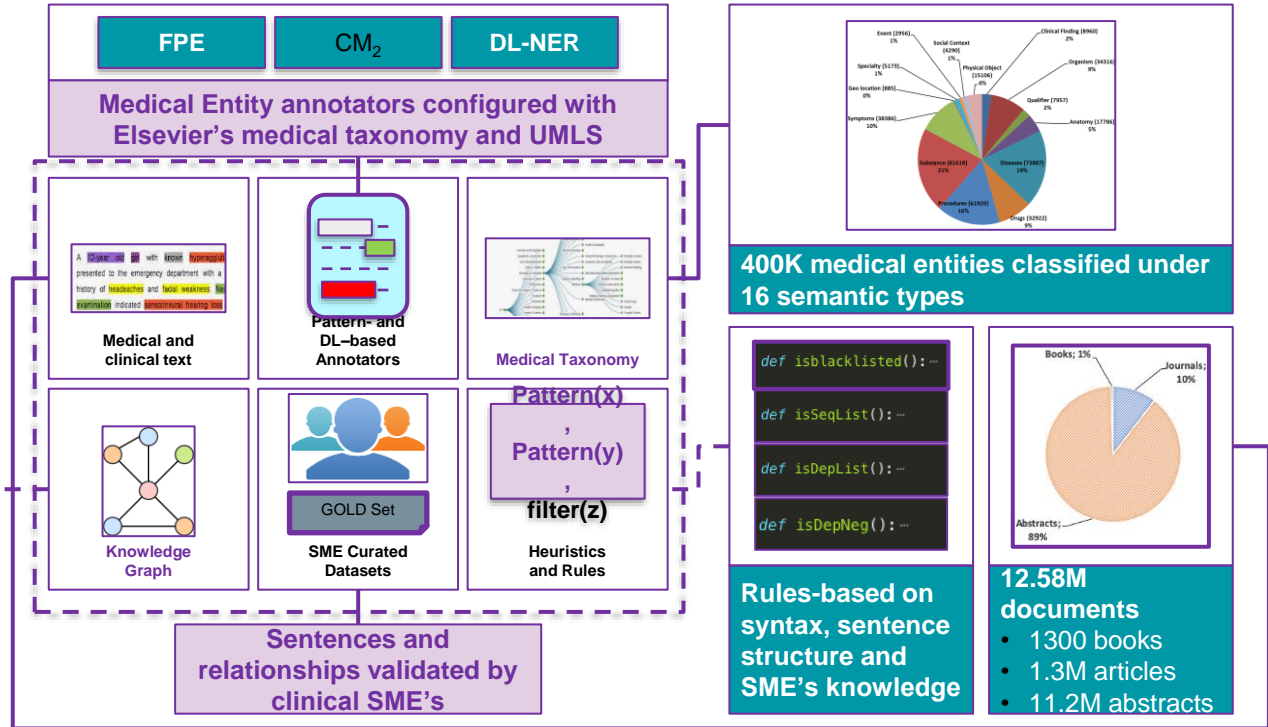
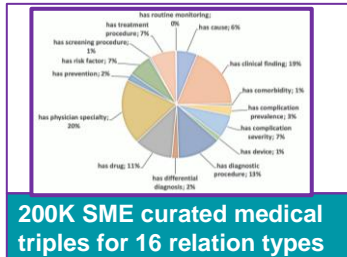


Language	# Medical name labels	# Synonym labels
English	400K	1M
Spanish	200K	400K
German	178K	167K
French	111K	210K
Brazilian Portuguese	150K	111K

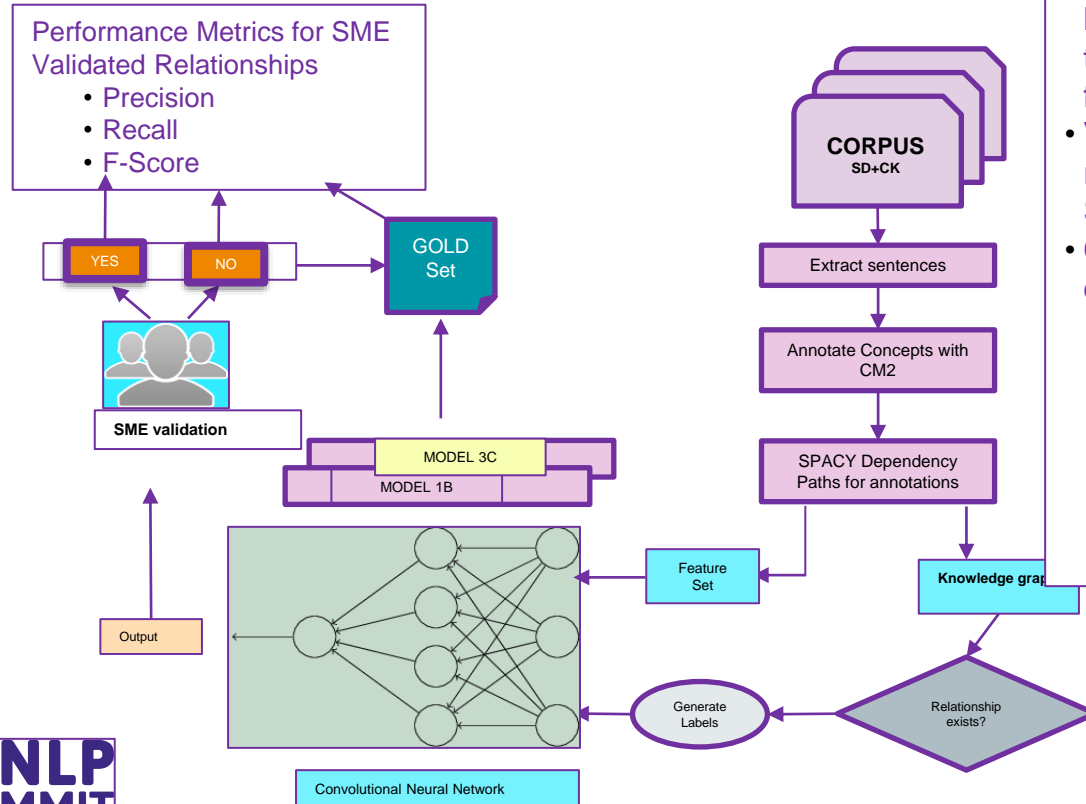


# Knowledge Extraction From “Noisy” Content

- Mining semantic patterns from our taxonomies
- Improving the quality of diverse data sets and annotations and search, extracting and identifying novel relationships for CDS support.
- Shared taxonomy and ontology used for breaking silos, connecting content, peoples, and different assets.



# Triple Extraction: Symptoms, Treatment, Diagnostic Procedure, and Drug Relationships from Medical Content at Scale



- Modeling workflow utilizes weakly supervised learning via convolutional neural networks (CNN) to predict symptom to disease relationships from free text content
- Validation done on a Gold Set in which relationships are assessed as true or false by SMEs.
- Connecting products and improving discoverability at Point-of-Care.

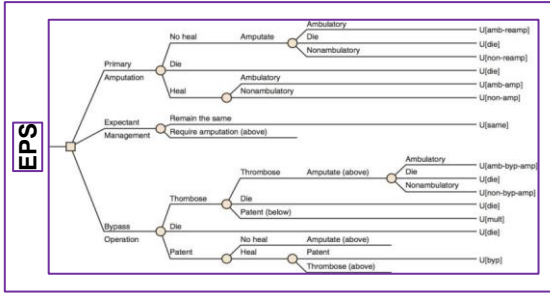
Add value by:

- Improving accuracy of treatment
- Reducing patient cost of unnecessary tests/treatment
- Reduced malpractice litigation

	Number of Documents
Books	1,320
Journals	1,317,321
Abstracts (Medline)	1,322,484
Total	2,641,125

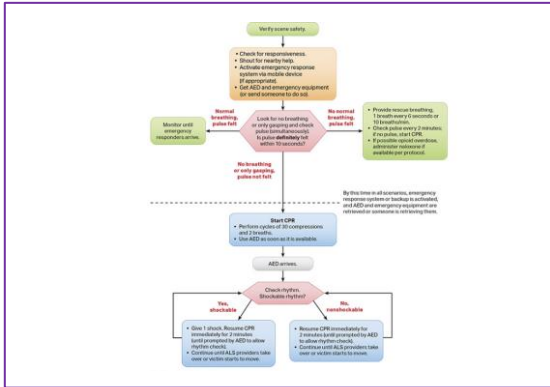
> 4.2 million clinical relationships extracted from unique evidence

# Automating Extraction of Medical Decision Trees Via Computer Vision



Flowchart or decision tree corpus is **heterogenous**

Different shapes, flow types, formats, colors,  
resolution, etc.



We developed a general, **computer vision-based AI system**

for **flowchart deconvolution** to be used at the  
**point-of-care**  
and in **education**



# Empowering Clinicians and Students TO Discover Content

- Using deep learning and data indexing techniques to develop **Topic Pages** that **are indexable and searchable on Google** for our customers
- Provide definitions and related terms for each concept
- Provide synoptic snippets of high-quality, relevant, peer-reviewed content
- Offer hundreds of thousands of comprehensive and topical web pages
- Free to access through our Elsevier Platforms

The screenshot shows a ClinicalKey Topic Page for 'Rheumatoid arthritis'. The page is divided into several sections. At the top, there is a definition of RA and a list of related terms. Below this, there are links to 'Add to Mendeley', 'Download as PDF', and 'Set alert'. The main content area is split into two columns. The left column contains a 'Clinical Overview' and 'Diagnostic Procedures' section, which lists primary diagnostic tools such as history and physical examination, laboratory testing, and the 2010 American College of Rheumatology/European League Against Rheumatism classification criteria. The right column contains a 'Guideline' section with an 'Overview' that discusses the purpose of the guideline and mentions NICE's COVID-19 rapid guideline on rheumatological autoimmune, inflammatory and metabolic bone disorders.



# Improving the User Search Experience

## Snippet evaluation

- Snippet images were reviewed by semantic group: Symptoms/findings, disease, drug, procedure
- Ranking system generated for image snippets

## Validating and Labelling definitions sentences

- Sentences prioritized by token
- Semantic patterns identified in sentence structure
- Definitions if multiple for a term were classified as preferred (Yes) or not preferred (No)
- 5,000+ definitions analyzed

## Images for Diffuse dermatitis



Fig 1 Clinical presentation on admission. Coalescent erythematous scaly papules in perioral area and periorbital edema are noted.

Lovett, Audrey, MD: Diffuse dermatitis: An unexpected initial presentation of cystic fibrosis. From: Journal of the



Fig 2 Diffuse erythematous scaly papules on trunk.

Lovett, Audrey, MD: Diffuse dermatitis: An unexpected initial presentation of cystic fibrosis. From: Journal of the American Academy of Dermatology.

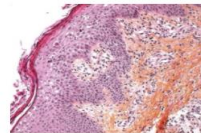


Fig 3 Specimen from first biopsy showing acanthosis, confluent parakeratosis, mild spongiosis, and mild lymphohistiocytic perivascular infiltrate with a few eosinophils. (Hematoxylin-eosin stain; original magnification:  $\times 20$ .)

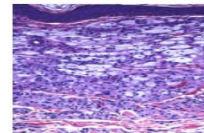


Fig 5 Later biopsy specimens showing predominately foamy histiocytes with rare Touton giant cells. (Original magnification:  $\times 10$ .)

Satter, Elizabeth K., MD, MPH: Diffuse xanthogranulomatous dermatitis and systemic

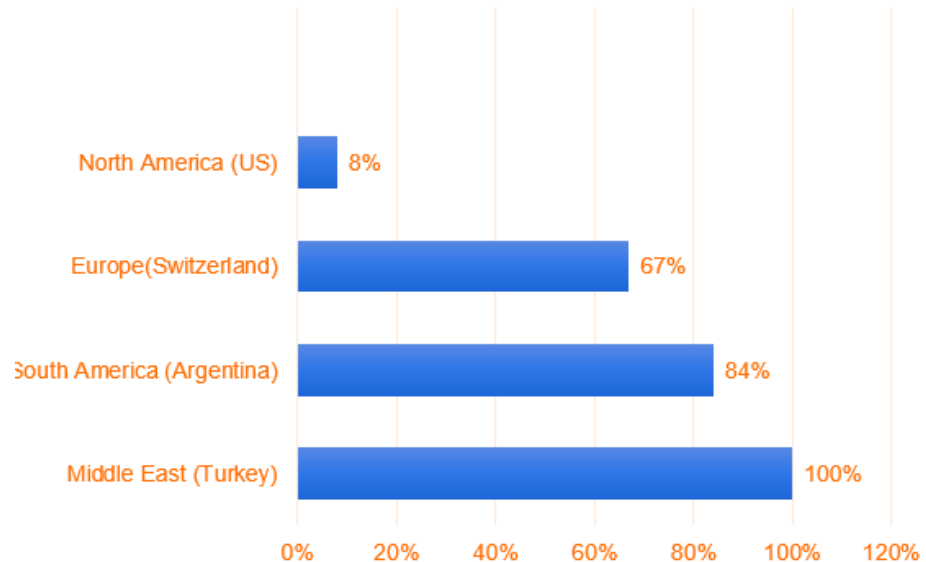
sentence	label	probab	tokens	concept	imul	Preferred
Constipation is defined symptomatically as the infrequent passage of hard stools, straining while passing a stool, or pain associated with the passage of a hard stool. Contact lens wear is another form of corneal microtrauma that seems to be associated with keratoconus.	TRUE	0.89073	0	Constipation	2791117	Y
Febrile seizures (FSs) are the most common type of pediatric seizure and occur in patients between 6 months and 5 years of age associated with fever but without intracranial infection or defined cause.	TRUE	0.58857	0	Contact lens	2791118	N
Febrile seizures are the most commonly reported neurologic adverse event after measles vaccination.268,726,7298733 Fever from any source lowers the threshold for seizures, and a febrile seizure is not a sign of central nervous system (CNS) infection or disease.	TRUE	0.77412	0	Febrile seizures	2791122	Y
A cosmetic is any substance applied to the body for cleansing, beautifying, promoting attractiveness or altering the appearance.	TRUE	0.57132	0	Febrile seizures	2791122	N
Unlabelled boxCosmeticsâCA cosmetic is a substance applied to the body for cleansing, to promote attractiveness or to alter the appearance.	TRUE	0.92267	1	cosmetic	2791125	Y
Tetracosactide is a synthetic peptide analogue which consists of the active N-terminal amino acids 1â24 of the ACTH molecule.	TRUE	0.90877	2	cosmetic	2791125	N
Craniosynostosis refers to the premature closure of one or more cranial sutures, which may occur during the prenatal period, early infancy, or childhood.	TRUE	0.60257	0	Tetracosactide	2791126	Y
Craniosynostosis is premature fusion of one or more cranial vault sutures with resultant skull deformity.	TRUE	0.81027	0	Craniosynostosis	2791127	N
Craniosynostosis is a congenital disorder that occurs due to premature intrauterine fusion of one or more of the six cranial sutures leading to an abnormal shape of skull.	TRUE	0.7404	0	Craniosynostosis	2791127	N
	TRUE	0.94814	0	Craniosynostosis	2791127	Y



# Most North American Guidelines are published in journals; Most international guidelines are published on websites

- Guidelines that are published in medical journals are usually indexed in literature databases such as MEDLINE
- A small but important proportion of North-American guidelines are published on websites rather than indexed MEDLINE journals
- A large proportion of international guidelines are not indexed on MEDLINE

## Proportion of Non-Indexed Guidelines by Region



<https://smw.ch/article/doi/smw.2019.20134>

<https://implementationscience.biomedcentral.com/articles/10.1186/1748-5908-3-43>

<https://www.dovepress.com/assessing-the-quality-of-clinical-practice-guidelines-in-the-middle-ea-peer-reviewed-fulltext-article-JMDH#t0002>

# Smart Surveillance BOT for Mining Dynamic

## ❑ Problem Statement

- Background: As the amount of content to review expands the importance of a smart literature surveillance BOT also increases.
- Bot provides dynamic streamlining by connecting a central repository guidelines that are easily searchable.

## ❑ Objective

- Creation of a surveillance BOT which can be run autonomously to gather dynamic changes in guidelines to improve evidence curation and decision support at the point of care.
- Empowering our clinician customers to better find how to diagnose and how to treat their patients.

Prototype for indexing and surfacing synoptic content across multiple guidelines

Search for condition or disease (peripheral artery disease)

peripheral a

Submit

- Peripheral angiopathy
- Peripheral anterior synechiae
- Peripheral arterial occlusive disease
- Peripheral arterial disease
- Peripheral arteriosclerosis
- Peripheral ameloblastoma

Explore related terms between concepts

Available mappings:  
ICD10 CM| Peripheral Artery Disease, unspecified| I73.9  
SNOMED-CT| Peripheral Artery Disease| 840580004

Domain/specialty (Oncology, cardiology, nephrology, etc...) Search

Select guideline type ☐ Diagnostic ☒ Therapeutic ☐ Preventative

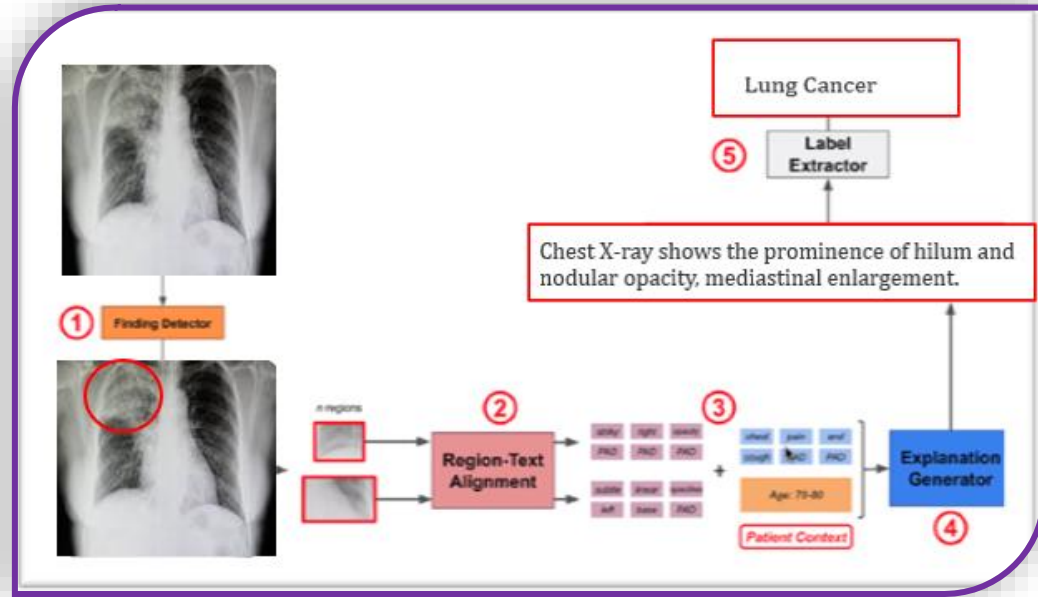
Language filter (EN, ES, FR, DE, ZN, etc...) Search



- There is a need to be able to provide predictive analytics at the point-of-care to improve patient outcomes by providing ***timely detection of deleterious signs*** in the hospital setting.

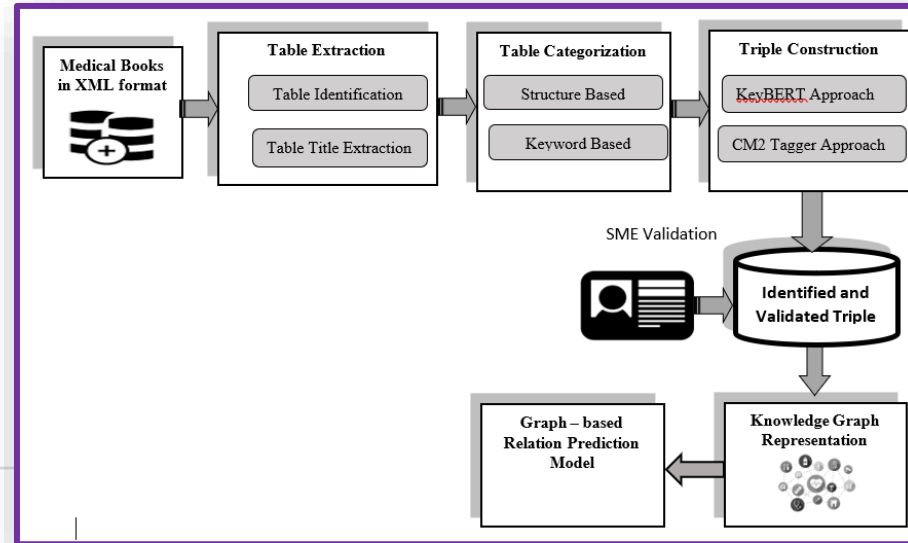
## Objective

1. Develop a predictive model for gaining insight into the patient journey and improving accuracy of diagnosis while decreasing the time to diagnose.
2. Algorithmically identify which features can be indicative of disease or prognostic outcome.
3. Provide support at the point of care.



# Semantic Table Extraction

- As the amount of scientific literature continues to grow exponentially, so does the need and time to review vast amount of content increase for clinicians and medical authors.
- Goal: Develop and implement tool aiding in knowledge extraction, in tabular format for purposes of generated synoptic content in graph form which can locate and identify relevant data by semantic category and prepare it for further processing and transformation.
- Benefits: Are many, improved authoring of content, information retrieval, decision support, and business intelligence.



# Semantic Table Extraction Pipeline

18.60 Causes of diabetes insipidus
<b>Cranial</b>
<b>Structural hypothalamic or high stalk lesion</b>
<ul style="list-style-type: none"> <li>See Box 18.54</li> </ul>
<b>Idiopathic</b>
<b>Genetic defect</b>
<ul style="list-style-type: none"> <li>Dominant (AVP gene mutation)</li> <li>Recessive (DIDMOAD syndrome – association of diabetes insipidus with diabetes mellitus, optic atrophy, deafness)</li> </ul>
<b>Nephrogenic</b>
<b>Genetic defect</b>
<ul style="list-style-type: none"> <li>V2 receptor mutation</li> <li>Aquaporin-2 mutation</li> <li>Cystinosis</li> </ul>
<b>Metabolic abnormality</b>
<ul style="list-style-type: none"> <li>Hypokalaemia</li> <li>Hypercalcaemia</li> </ul>
<b>Drug therapy</b>
<ul style="list-style-type: none"> <li>Lithium</li> <li>Demeclocycline</li> </ul>
<b>Poisoning</b>
<ul style="list-style-type: none"> <li>Heavy metals</li> </ul>
<b>Chronic kidney disease</b>
<ul style="list-style-type: none"> <li>Polycystic kidney disease</li> <li>Sickle-cell anaemia</li> <li>Infiltrative disease</li> </ul>

Table image from Medical  
Book PDF

1. Prepend a column header to each concept
2. Detect row header – if any, prepend it with each concept from row
3. Generate a triple with {table\_keywords, Column\_header, Concept\_Value}

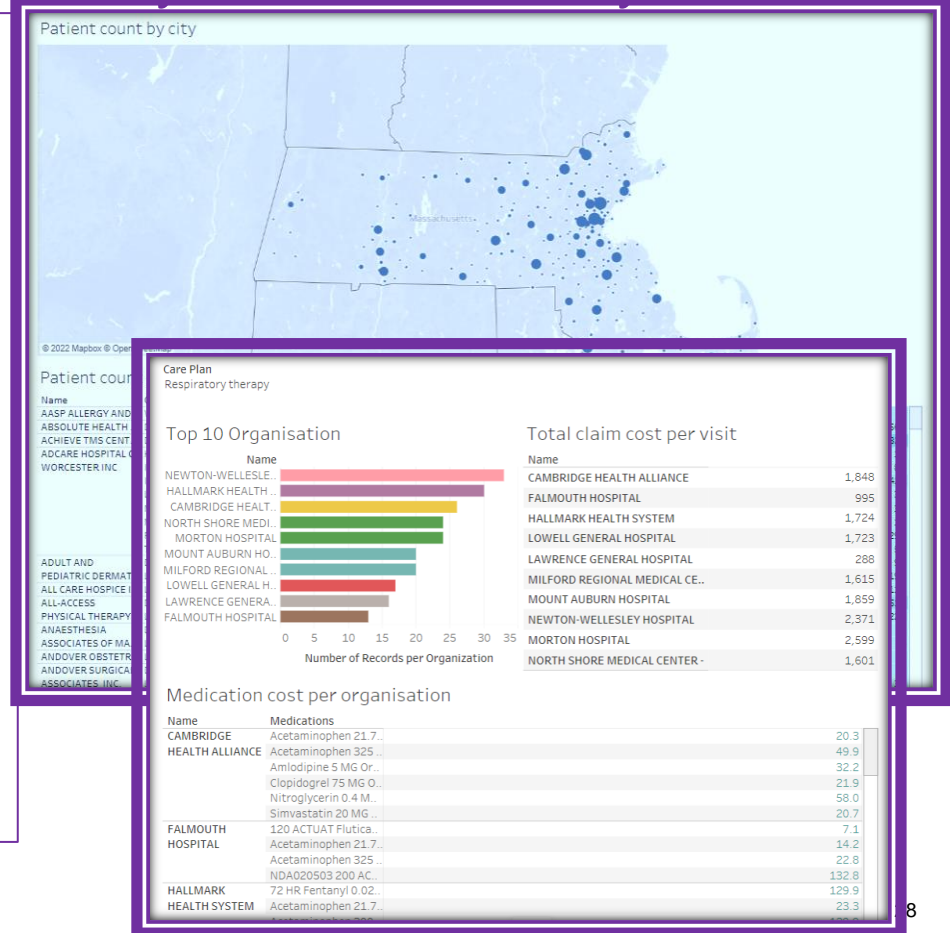
Steps to extract data

index	table_id	Table_name	Keywords	Predicate	Concept_list
0	tit0015	Root causes of diagnostic error in studies	['root causes', 'diagnostic error studies']	Error category - No fault	'Examples - Missing information']
1	tit0015	Root causes of diagnostic error in studies	['root causes', 'diagnostic error studies']	Error category - No fault	Examples - Unusual presentation of a disease',
2	tit0015	Root causes of diagnostic error in studies	['root causes', 'diagnostic error studies']	Error category - System error	'Examples - Inadequate diagnostic support'
3	tit0015	Root causes of diagnostic error in studies	['root causes', 'diagnostic error studies']	Error category - System error	'Examples - Results not available'
4	tit0015	Root causes of diagnostic error in studies	['root causes', 'diagnostic error studies']	Error category - System error	'Examples - Error-prone processes'
5	tit0015	Root causes of diagnostic error in studies	['root causes', 'diagnostic error studies']	Error category - System error	Examples - Poor supervision of inexperienced staff
6	tit0015	Root causes of diagnostic error in studies	['root causes', 'diagnostic error studies']	Error category - System error	'Examples - Poor team communication'
7	tit0015	Root causes of diagnostic error in studies	['root causes', 'diagnostic error studies']	Error category - Human cognitive error	'Examples - Inadequate data-gathering'
8	tit0015	Root causes of diagnostic error in studies	['root causes', 'diagnostic error studies']	Error category - Human cognitive	'Examples - Errors in reasoning']

Generated Triples

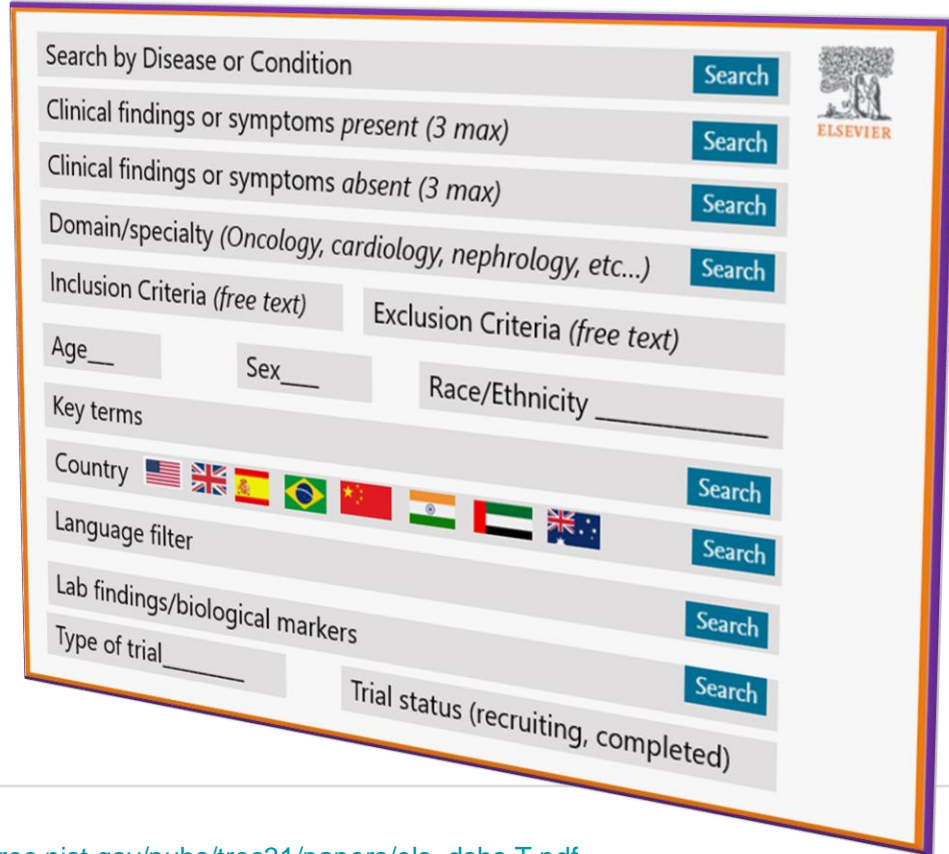
# Predictive Analytics Using Pseudonymized Data & Synthetic Data

- Displays adherence of products in relation to data sets
- Facility, location, date, condition
- Identifying Plan Adherence
- Automating Reporting
- Viewer advantage of interactive Tableau dashboard vs. manually reviewing PDF's.
  - Patient readmission
  - Care team utilization
  - Nurse staffing
  - Hospital flow simulation etc.



# Automated Patient-to-Trial Finder to Improve Access to Care

- Clinical trials are the cornerstone of evidence-based medicine, ensuring the availability of safe and effective treatments by studying their effects on human subjects.
- Matching eligible patients to clinical trials is essential for achieving statistically significant results; however, the recruitment of patients represents a bottleneck in clinical research.
- To better understand the limitations of our approach, we performed a manual evaluation of 600 topic-trial recommendations, demonstrating the importance of accounting for the trial exclusion criteria in retrieval.
- SOLUTION: Developed a prototype via application of transformer embeddings using MiniLMmodel.



The image shows a screenshot of a clinical trial search interface. It features several search filters and buttons, each with a blue 'Search' button to its right. The filters include:

- Search by Disease or Condition
- Clinical findings or symptoms *present* (3 max)
- Clinical findings or symptoms *absent* (3 max)
- Domain/specialty (*Oncology, cardiology, nephrology, etc...*)
- Inclusion Criteria (*free text*)
- Exclusion Criteria (*free text*)
- Age \_\_\_\_
- Sex \_\_\_\_
- Race/Ethnicity \_\_\_\_
- Key terms
- Country (with flags for USA, UK, Spain, Brazil, China, India, UAE, Australia)
- Language filter
- Lab findings/biological markers
- Type of trial \_\_\_\_
- Trial status (recruiting, completed)

The interface is framed by a purple border. In the top right corner, there is a logo for Elsevier, featuring a tree and the word 'ELSEVIER'.

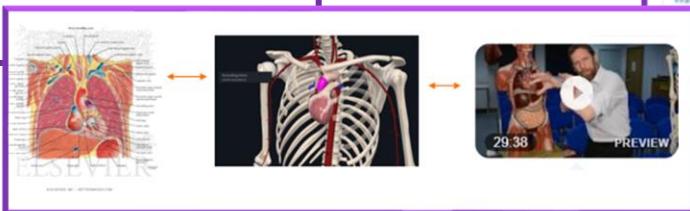
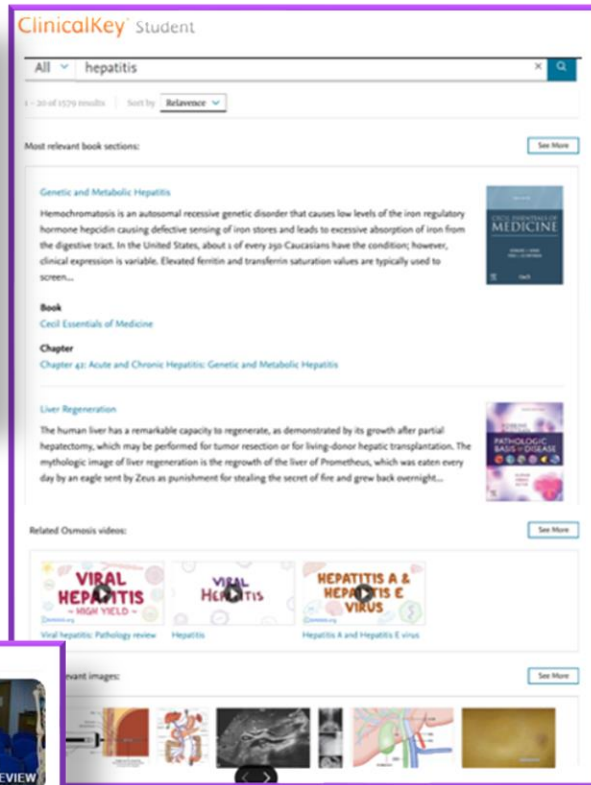
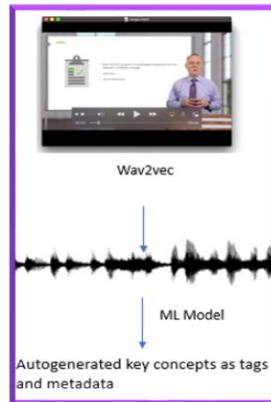
# Data Science Improving Educational Offerings



# Automated Transcript Generation and Metadata Tagging for Linking Elsevier Health Products for Medical Students

Elia Lima-Walton,  
MD  
Elsevier  
March 27, 2023

- Improved content indexation and interlinking between transcripts and clinical vignettes
- Facilitating navigation and interoperability across education products
- Used to create model which prioritizing key terms/meta data to be surfaced
- Additionally:
  - Automated annotation and parsing
  - 2D/3D image cross walks created
  - Automated prioritization of key to support various Products





# Cultural Sensitivity Prototype for Nursing Health Education

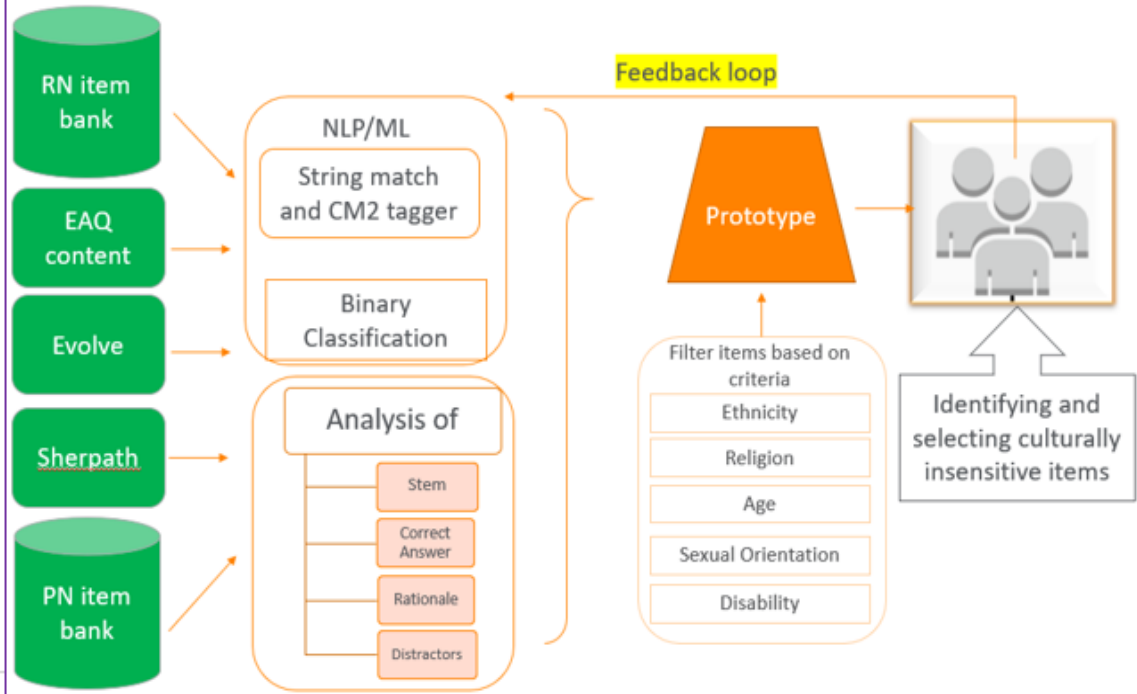
## Objective

1. Investigate which items are culturally insensitive in nursing practice test items.
2. Provide analytics of topic area and banks containing culturally insensitive items.
3. Build a prototype using ML, NLP, & Analytics which identifies items from nursing banks and remediation material that is culturally insensitive.

## Key Impact

Text extraction, ingestion, and binary classification to feed into algorithm and built prototype for NHE support.

## NHE Cultural Sensitivity Prototype





# Supporting Patient, Student, and Researcher's Journey

“At the heart of improving healthcare is effectively applying outcomes-driven analytics across the care continuum. I’m excited about the tremendous opportunity to combine the market-leading evidence based content and clinical reference and workflow solutions of Elsevier Health with our differentiated data and analytics capabilities to support healthcare providers in delivering better outcomes for patients.”

**Josh Schoeller**, *President Clinical Solutions, Chief Executive Officer of the Health Care business of LexisNexis® Risk Solutions*





# Thank you

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