

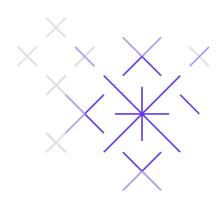
Leveraging Healthcare NLP Models in Regulatory Grade Oncology Data Curation



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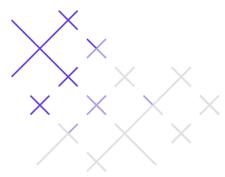


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Bringing clarity to cancer

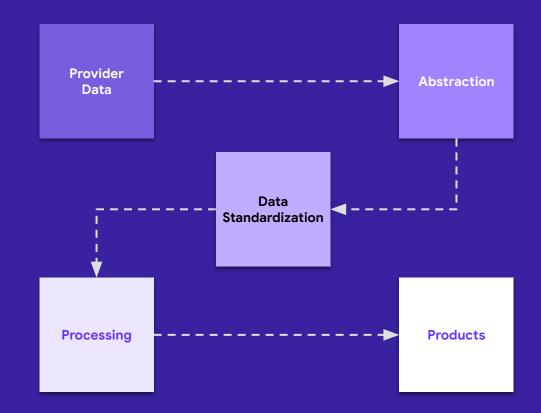
At COTA, we are working to ensure that everyone touched by cancer has a clear path to the right care.





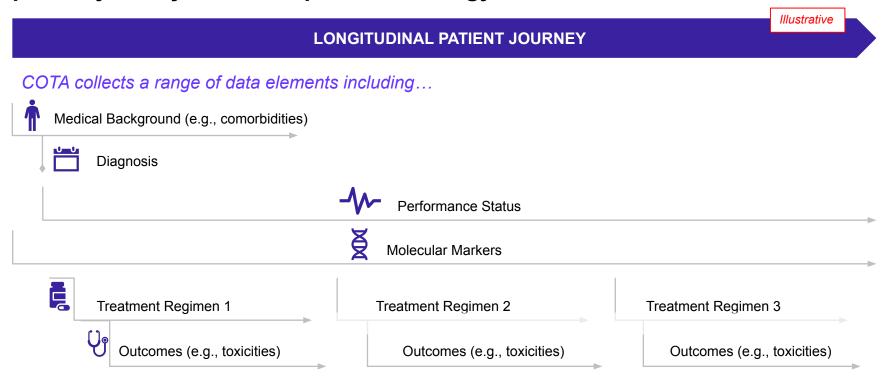
How we do it

COTA organizes fragmented, often hidden data from the real world, through tech-assisted and enhanced human abstraction.





The COTA platform provides a clinically comprehensive and longitudinal patient journey for each specific oncology cohort





RWD Clinical Data Abstraction Challenges

Challenge

Options



Manual
Clinical Data
Abstraction

High clinical accuracy



Structured EMR Data

Fragmented, non-standard, maintenance heavy



Search / Text Processing

Rules based, somewhat unpredictable



Generic Healthcare NLP Models

Useful in some contexts



Domain specific Pretrained Models

Great for a start, but needs tuning



NLP Use cases explored



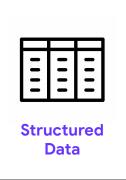
Highlighting relevant text for abstraction

Extraction of data for review / usage



Model Tuning & Active Learning

Annotation Lab + Feedback loop through Abstractors

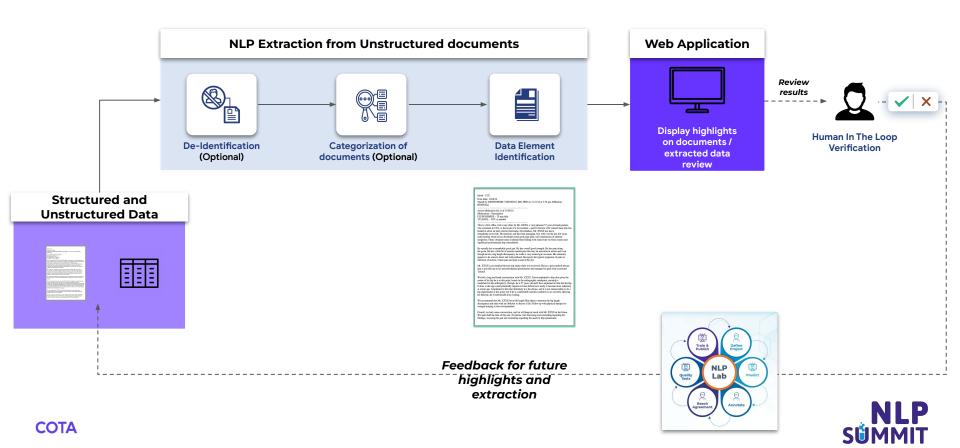


Extract normalized data for review from free text fields

De identify free text / structured text



Model Enhancements & Active Learning Process



Adverse Events

 Any unfavorable and unintended sign (including an abnormal laboratory finding), symptom, or disease temporally associated with the use of a medical treatment or procedure. Also called a toxicity.

B-Symptoms

- Fever Temperature >38°C (>100.4°F)
- Weight loss Unexplained loss of >10 percent of body weight over the past six months prior to diagnosis
- Sweats The presence of drenching night sweats



Adverse Events

SUMMARY: In summary, Mr.

has CLL. He has been receiving fludarabine/rituximab. He has tolerated the first four cycles well. The red cell and platelet counts are normal. He remains fatigued.

The WBC has decreased; he has mild

neutropenia +ADVERSE EVENT

We will defer therapy today. I will see him again in 2 weeks; we plan to proceed with FR likely at a reduced dose when counts are adequate.

We again discussed the treatment plan and possible side effects/toxicity.

B-Symptoms

IMPRESSIONS:

1. Extensive marrow infiltration for which the differential diagnosis reported includes "myeloproliferative disorder (chronic anemia, myelofibrosis, leukemia or lymphoma) or marrow infiltrating metastatic disease"

Weight loss, chills, and night sweats\n3.
 +BSYMPTOMS

Bilateral lower extremity pain



B-Symptoms Model Details

- Detect Lymphoma, B-Symptoms, Dates of Assessment
- Initial accuracy somewhat lower for practical use cases
- Clinical expertise required to train the model
- Additional training increased accuracy by 20%
- Increased accuracy can be achieved with more training and data points



Challenges

Side effects of treatment explained at length including cytopenias, neuropathy, fever, and fatigue.

+BSYMPTOMS +ADVERSE EVENT -ADVERSE EVENT

Hem/Onc

Diagnosis Non-Hodgkin's lymphoma +ADVERSE EVENT : Stage at diagnosis: IVB

Review Of Systems General: As above. Head, Eyes: Negative. ENT: Negative. Cardiovascular: Negative +ADVERSE EVENT Pulmonary: Negative +ADVERSE EVENT GI: Negative -ADVERSE EVENT GU: Negative -ADVERSE EVENT Musculoskeletal: Negative +ADVERSE EVENT Dermatologic: Negative +ADVERSE EVENT Neurologic: Negative -ADVERSE EVENT Extremities: Negative -ADVERSE EVENT



Future Use: Outcomes

- An outcome is a measure of the overall response to systemic treatment
- Ranges from CR (Complete Response) to Progression/Relapse

5/1/2019 PET scan—<u>favorable response to therapy</u> v PR ecreased size and <u>resolution of activity</u> in mesenteric lymphadenopathy and decreased activity and bilateral lung hila. Mesenteric nodal mass 20 x 17 mm, SUV 1.6, previously 62 x 44 mm. Nodular airspace opacity in the lingula of the left lung stable in size, increase activity. Lingular nodule currently 19 x 11 mm, SUV 3.4.



Summary

Challenges / Learning:

- Challenges persist when trying to identify healthcare related data within unstructured text
 - Oncology is even more challenging given that treatments are personalized
- Creating completely new models from scratch to solve data extraction is possible but requires in-house expertise in understanding how best to operationalize the models
 - Data for training is often sparse in Oncology

Positives:

- NLP models can help with extraction of data at scale, with quality, as long as there's a good review
 / learning process
- Time savings for Human Abstraction can be realized in targeted data extraction scenarios by highlighting as well as review of extracted data.
- Models can be improved with data labeling and active learning
 - Tooling is essential for improving accuracy of models
- Multiple models and pipelines are often needed, but there are lots of choices out there.







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