

Hidden Network Monitoring

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Overview – General Statements & Hypothesis

- Analysis of Publicly Available Information (PAI) can offer valuable information for understanding nonproliferation and weapons development. AI can help this analysis.
- Specifically Natural Language Processing (NLP) has advanced from sentiment analysis to understanding the intent of the communicator (psychology and health).
- Expert Knowledge aka Human-in-the-loop can compensate sparse and/or sporadic events in certain domains (e.g., directed scientific discovery).
- Using NLP with domain knowledge of SMEs; resulting sequences and clustering of queries can help identify signatures indicating some probabilistic measure of a user's intent (academic searches vs. searches for the development of a weapons program).



Overview – Technique and Domain

General

 AI analysis of requests to Publicly Available Information (PAI) to help determine a user's "intent".

Specific

 Use Natural Language Processing on queries to Publicly Available repository of Nuclear Science Data at DOE national lab to reveal *questionable* intent.

"The use of publicly available data and analysis is effective at identifying high-risk nuclear trade." "Machine learning tools can be used to dramatically enhance data analysis in terms of both speed and quality."

*E. Moniz: Signals in the Noise – PREVENTING NUCLEAR PROLIFERATION WITH MACHINE LEARNING & PUBLICLY AVAILABLE INFORMATION C4STS / NTI 2021



Overview – Technical Approach

- Backend Data Collection
 - Operate on data while in the network and in transit
 - Collect, clean, filter, bin and hand-off to Analysis
 - Possible use includes preprocessing and some analysis
- NLP
 - Short text clustering (queries which not grammatically correct)
 - building good representation
 - Early classification
 - through multi-instance learning
 - Intent detection (probabilistically predictive / through evidence found through a sequence of queries)
- SME
 - Obtain cross-complex expert knowledge regarding query relevance to input into system



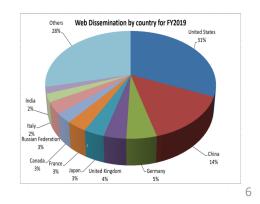
Goals - General

- Early Detection of a Weapons Program or Proliferation Activity
 - Gathering of Foundational Science (as far upstream as possible)
 - e.g., Literature searches are revealing
 - Use of collective SME knowledge to aid in detection analysis
- Monitoring of PAI containing valuable information
- Creation of a silent data collection system that supports monitoring and contributes to the analysis on the fly and in near real-time



Goals - Additional Community Benefits

- A network-based filtering / cleaning / analysis tool
 - Efficient
 - Rapid Response
 - Federation
 - Clandestine
- Contributing to a Self Supervised approach toward early classification problems
- Domain Aware NLP "human-in-loop"
 - Greater understanding on the PAI and its users
- Monitoring of relevant PAI data repository
 - Simple anomalous behavior
 - Domain relevant statistics
- Code that is adjustable, reusable and scalable
 - Modular and not data specific

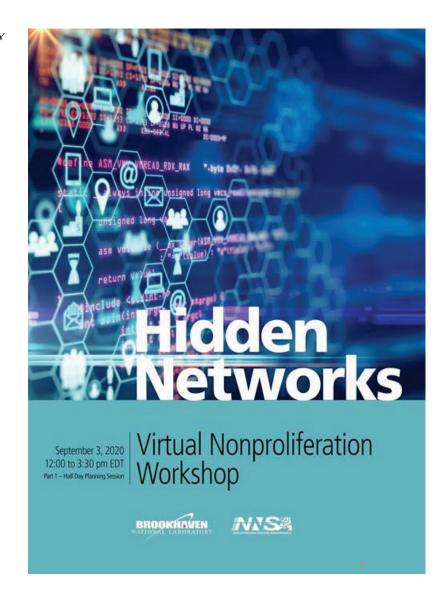


Nonproliferation (human-in-loop)

 Literature review conducted to identify materials of interest and results distributed to points of contact for feedback

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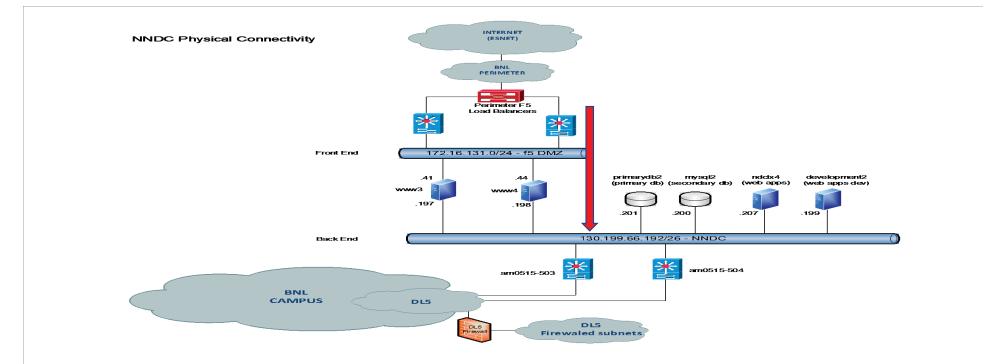
- Log review Eyeball/simple script analysis to identify clustering/significant IPs
- Following up on other recommendations from the September workshop
- Provide expert knowledge regard significance of sequences for AI tuning





Collection progress - Network

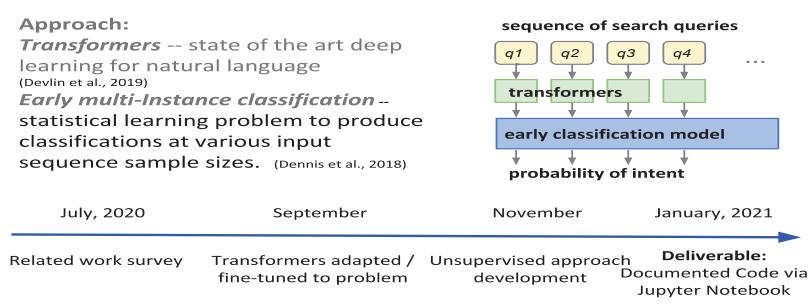
Configured and installed monitoring node in production networking architecture Packets captured and NNDC logs reconstructed





Analysis

Intent Detection from Search Sequences



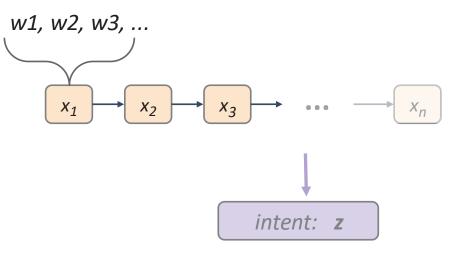


Progress - Analysis

- Analysis Code Nutshell
 - Creates NL Higher Dimension Representations with primary keys then sets into sized vector. Creates clustering of significant sequences.
 - Queries pass through transformers (RoBERTa-base) to be embedded and then the sequence of embeddings (produced from the transformers) are fed to an RNN-LSTM to do the classification.

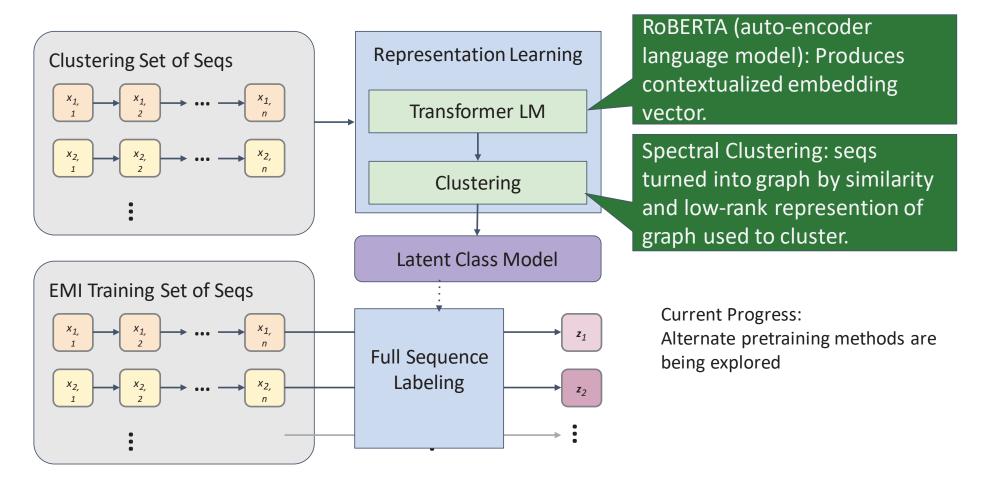
Each query is a sequence of *keywords* or *natural language*

Each query is part of a sequence of multiple queries with an intended target.



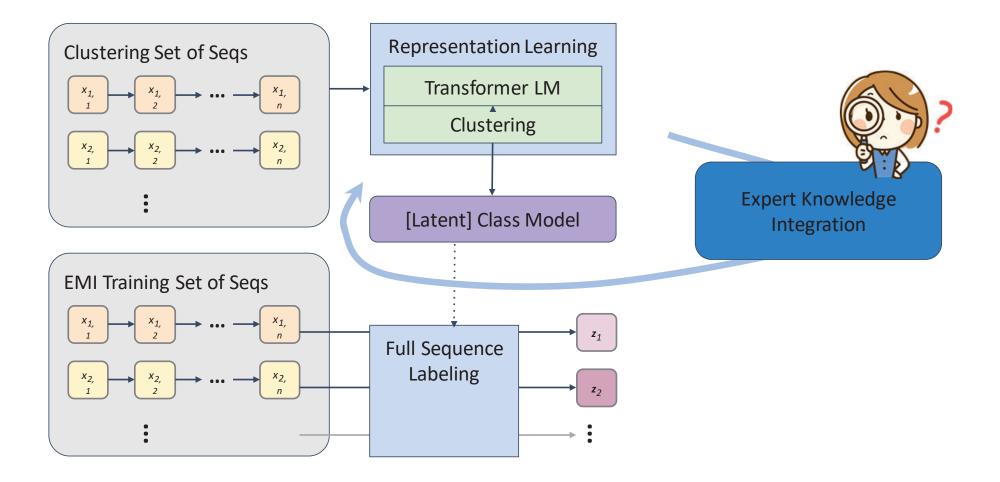


Self Supervised Approach – Early Classification



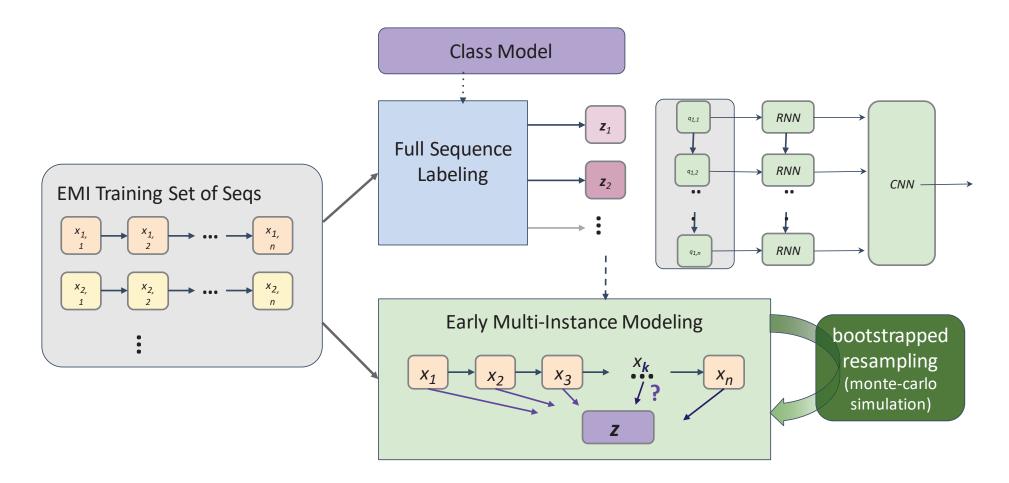


Fine Tuning Early Classification with Expert Knowledge





Prediction Confidence Evaluation (95%)





Clustering – Sanity Check

Reviewing the clustering results by looking at queries from 5 sequences from each cluster In [258]: records = queryData_1.data["values"] for cluster_num in range(num_clusters): random caed(d3) Upto first 10 queries of 5 random sequences from cluster 0 (separated by ;): [(1, 'Barium; 116Barium; 116 Barium; Barium 116; Barium; Barium, production'), (2, '32 IMME; 32 IMME; IMME'), (3, 'helios,gas cell; helium, implanted; 3he, implanted'), (4, 'Slater approximation'), (5, 'beam energy, spin; spin')] [(1, 'Barium; 116Barium; Barium 116; Barium; Barium, production'), (2, '32 IMME; 32 IMME; IMME'), (3, 'helios,gas cell; helium, implanted; 3he, implanted'), (4, 'Slater approximation'), (5, 'beam energy, spin; spin')]



Next steps

- Establish information of interest
- Establish baseline on historic data
- Forward information directly to the algorithm streaming logs
- It may be possible to correlate multiple queries/responses to extract such information, but this moves from filtering to more intelligent processing and it remains to be decided where and how this should take place.
- Implementing a more state-of-the-art machine learning technique to do the classification
- Integrating expert knowledge to improve the classes/labeling/clusters
- Iterative evaluation integrating additional data
 - Accuracy / performance
 - Metadata
- Process historic data