Development of a Lexicon for Pain

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Pain

- Pain has been an active area of research, especially since the onset of the crisis of opioid use in the United States [1]
- Pain is known to have a strong relationship with emotions, which can lead to damaging consequences - worsened for people suffering with persistent pain [2]
- Apart from research, it has been of interest to the general population as well - google trends
Google trends - search term

- Pain: Medical condition
- Fever: Medical condition
- Cough: Disease

Interest over time

Worldwide  2004 – present  All categories  Web Search

Note: The graph shows the interest in search terms related to pain, fever, and cough over time, with a significant increase from 2020 onwards.
Publications for “pain” - Web of Science
Why am I interested in pain?

- The term “pain” presents a unique natural language processing (NLP) problem
  - subjective nature
  - ambiguous description
- Pain can refer to physical distress, or existential suffering, and sometimes metaphorical in phrases such as “for being a pain” [3]
- Within the biomedical context, it will most likely be the former two.

**Purpose of this study** - Develop an English lexicon of pain terms for use in NLP tasks
Exploration of Pain: Data Sources

1. Electronic Health Records
   a. CRIS
   b. MIMIC-III

2. Social Media
   a. Reddit
   b. Twitter
Electronic Health Records

CRIS database

- Anonymised version of EHR data from the South London and Maudsley NHS Foundation Trust (SLaM).
- Contains ~30 million event notes and correspondence letters, with an average of ~90 documents per patient
- Data extraction:
  - A SQL query was run on the attachments table of the CRIS database, and random 50 documents that contained the keyword 'pain' (both upper and lower case) were extracted.
Electronic Health Records

MIMIC-III

- Medical Information Mart for Intensive Care (MIMIC) is an openly available database which was developed by the Massachusetts Institute of Technology (MIT).
- De-identified healthcare data from the critical care units
- Discharge summaries and clinician progress notes average to about 3,900 words per document
- Data extraction:
  - A SQL query was run on the note-events table, and random 50 documents containing the keyword ‘pain’ (both upper and lower case) were extracted.
Social Media

Reddit

- A popular online community which supports unidentifiable accounts to allow users to post anonymously.
- Provides sub communities for people to discuss topics of shared interest (SubReddits)
- Chronic pain subreddit community - r/ChronicPain
- Data extraction:
  - Python package - PRAW
  - All posts from ChronicPain subreddit community (7,700 posts)
  - Random 50 used for analysis
Social Media

Twitter

- An online micro-blogging platform with an enormous number of users who post short (280 characters or less) messages, referred to as “tweets”, on topics of interest.

- Data extraction:
  - Python package - tweepy
  - All tweets containing words “chronic pain” (7,701 posts)
  - Random 50 used for analysis
A note on ethics and data access

- While data from Reddit and Twitter are publicly available, applicable ethical research protocols proposed by Benton et al. (2017) were followed in this study [4].
- No identifiable user data was used, and any sensitive direct quotes were paraphrased.
- Data from Twitter is available through their API after approval of registration for access to this data, details of which can be found in their general guidelines and policies documentation [5].
- Data access information for CRIS [6] and MIMIC-III [7] are detailed on their respective websites.
Collocates for “pain”

<table>
<thead>
<tr>
<th>CRIS</th>
<th>MIMIC-III</th>
<th>Reddit</th>
<th>Twitter</th>
</tr>
</thead>
<tbody>
<tr>
<td>chronic</td>
<td>control</td>
<td>pain</td>
<td>agony</td>
</tr>
<tr>
<td>back</td>
<td>acute</td>
<td>about</td>
<td>amazingly</td>
</tr>
<tr>
<td>clinic</td>
<td>chronic</td>
<td>anyone</td>
<td>achieved</td>
</tr>
<tr>
<td>physical</td>
<td>assessment</td>
<td>back</td>
<td>american</td>
</tr>
<tr>
<td>health</td>
<td>plan</td>
<td>anything</td>
<td>body</td>
</tr>
</tbody>
</table>
# Common themes

<table>
<thead>
<tr>
<th>Source</th>
<th>Type of pain</th>
<th>Feelings/experiences asso. with pain</th>
<th>Medications and other measures</th>
<th>Related to body parts</th>
</tr>
</thead>
<tbody>
<tr>
<td>CRIS</td>
<td>in constant pain..</td>
<td>overwhelmed by chronic pain problems..</td>
<td>drugs to numb the pain..</td>
<td>chronic back pain..</td>
</tr>
<tr>
<td></td>
<td>ongoing pain..</td>
<td>fear of pain..</td>
<td>pain relief medication not controlling the pain..</td>
<td>chest pain..</td>
</tr>
<tr>
<td></td>
<td>pain was quite severe..</td>
<td>pain causing distress..</td>
<td>side effects from pain relief medication..</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>struggles with chronic pain..</td>
<td></td>
<td></td>
</tr>
<tr>
<td>MIMIC-III</td>
<td>severe pain..</td>
<td>-</td>
<td>PO as needed for pain..</td>
<td>chronic back pain..</td>
</tr>
<tr>
<td></td>
<td>atypical pain..</td>
<td></td>
<td>taking narcotic pain medication..</td>
<td>chest pain..</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>managed with IV pain medication..</td>
<td>abdominal pain..</td>
</tr>
<tr>
<td>Reddit</td>
<td>Sharp pain..</td>
<td>could be causing pain</td>
<td>helped my back pain..</td>
<td>chronic neck pain..</td>
</tr>
<tr>
<td></td>
<td>Widespread pain..</td>
<td>painful trips to kitchen</td>
<td></td>
<td>back pain..</td>
</tr>
<tr>
<td></td>
<td></td>
<td>am in the same painful position I was 3 months ago..</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Twitter</td>
<td>-</td>
<td>to live pain-free..</td>
<td>muscle painbuster amazing discount!</td>
<td>back pain</td>
</tr>
</tbody>
</table>
Conceptual diagram of pain
Building the Lexicon

Terms were generated from 3 different sources:

1. Literature
2. Ontologies
3. Embedding models
Pain-related words were harvested from three publications which made pain-related terms available in their supplemental materials:

1. A systematic review on application of NLP methods for symptom extraction from electronic patient-authored text (ePAT) [8].
2. A survey of biomedical literature based word embedding models [9].
3. A list of sign and symptom strings generated using NLP to meaningfully depict experiences of pain in patients [10].

These lists were cleaned by lowercasing all terms, and only keeping terms made up of 1 or 2 tokens.
Ontologies

Synonyms for pain were incorporated from three biomedical ontologies:

1. The Unified Medical Language System (UMLS) [11],
2. Systematized NOmenclature of MEdicine Clinical Terms (SNOMED-CT) [12],

UMLS contains concepts from SNOMED-CT and ICD-10, in addition to a number of other vocabularies.

From each, we extracted terms of up to 2 tokens that either matched “pain*”, were synonyms of pain, or described as child nodes of pain.
Embedding models

Embedding models [36, 37] using 8 different parameters and four different text sources were used to generate additional words similar to “pain”:

1. Two of the embedding models built using clinical text within MIMIC-II database.
2. Four embedding models built using clinical text within MIMIC-III (three using gensim implementation of word2vec and one using FastText.)
3. A model was built using word2vec over an SMI cohort from CRIS.
4. A publicly available model built on PubMed and PubMed Central (PMC) article texts

Only unigrams were included from the different models. Any duplicates were removed, and the remaining terms were added to the lexicon.
Lexicon coverage

<table>
<thead>
<tr>
<th>Lexicon source</th>
<th># of unique terms</th>
<th>Total # of terms</th>
</tr>
</thead>
<tbody>
<tr>
<td>Literature</td>
<td>218</td>
<td>241</td>
</tr>
<tr>
<td>Ontologies</td>
<td>291</td>
<td>523</td>
</tr>
<tr>
<td>Embeddings</td>
<td>68</td>
<td>171</td>
</tr>
</tbody>
</table>
Validation of Lexicon

- The final list was validated by two clinicians, with the final size of the lexicon being 382 terms (30% unigrams, 68% bigrams, and 1% trigrams).

- Further validation was conducted by comparing the lexicon to another ontology which consisted of some pain terms - the Experimental Factor Ontology [14]:
  - 70% of the terms within the pain lexicon were also in the Experimental Factor Ontology, with all pain-related terms being covered (such as pain, ache, cramp)
Some patterns

1. Anatomy
   <anatomy term> <pain term> such as chest pain, ear ache, head discomfort
   <pain term> <anatomy term> such as acute abdomen, aching muscles

1. Quality of pain
   <quality term> <pain term> such as burning pain, chronic pain
   <pain term> <quality term> such as pain burning, pain crushing

1. Quality of pain and anatomy
   <quality term> <anatomy term> <pain term> such as chronic back pain
Conclusion

- The final pain lexicon and the code to generate the embedding models is openly available on GitHub*
- The lexicon will also be added to other ontology collections such as BioPortal.
- This final lexicon will be used in downstream tasks such as building an NLP application to extract mentions of pain from clinical notes.

*https://github.com/jayachaturvedi/pain_lexicon
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Thank you!
Questions?
References

6 https://www.maudsleybrc.nihr.ac.uk/facilities/clinical-record-interactive-search-cris/
7 https://mimic.physionet.org/