



Progress Presentation

Sponsors: Focused Ultrasound Foundation

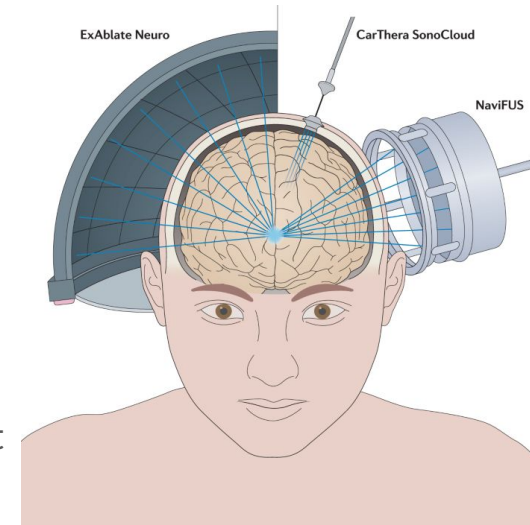
Faculty Mentor: Heman Shakeri

Student Team: Rose Eluvathingal Muttikkal, Abhishek Singh, Skye Jung, Reanna Panagides



Project Purpose and Background

- **Problem:** The current literature review process is timely, inefficient, and prone to error
- **Proposed solution:** Utilize *machine learning techniques* to enhance the efficiency of *classifying scientific literature* about *emerging applications* and *indications* of focused ultrasound therapies



Meng et al, 2021

Project Plan

Data format ✓
Excel to pandas
dataframe

EDA ✓
~480 labeled
articles

Data prep
Remove
irrelevant
characters

Train
SL approach
SBERT

Assess
Test data
FUSF feedback

Updates 1/26/24

```
0    441
1     48
Name: fus_related, dtype: int64
```

Sean's work

- Retrieve articles related to **ML/AI** in focused ultrasound or its related domains
- **PDF processor** that enables users to input any PDF article
- Article summarized → **ChatGPT's API**/analyzer that rates sentence relevance
- **Classification** of articles → FUS/Non-FUS, Supervised/Unsupervised)

```
prostate    6
UF          4
AD          3
bone mets   3
liver       3
ET, PD      2
obesity     2
stroke intracerebral hemorrhage 2
pd          2
DIPG        2
uf          2
tendon contracture 1
spinal cord injury 1
soft tissue tumors 1
ut adenomyos 1
Retained placenta 1
Ablation- other 1
OCD         1
PD          1
Pancreatic ca 1
lichen sclerosis 1
heart failure 1
GBM         1
ET,         1
cervicitis  1
Cervical cancer 1
atherosclerosis 1
ut adenomyosis 1
Name: ind_list, dtype: int64
```

Updates 2/9/24

SPRINT 2 - Model(s) Building

02/01

02/14

- Sean Meeting
 - Current usage of Sean's work?
 - Cost to host Sean's work?
- Obtaining summer data? (Samarth + Eric)
- SIEDS Abstract
 - <https://docs.google.com/document/d/1TsKA0zhi0HTdYaJ6m2Ui5fgWGKMLzGtCrjeQSBfLA6I/edit?usp=sharing>
 - Co-author?
- Currently working on...
 - Running previous scripts locally → app.py, model file
 - Web scraping FUS website
 - Running BERT model on "test data"

Updates 2/21/24

- **Sprint 2 (Model Building) Complete**
 - Finished SIEDS Registration
 - Sean's work
 - Met with him
 - Got access to Sean's data (~90 labelled manuscripts)
 - Understood model development (limitations)
 - Ran BERT mode on "test data"
- **Started Sprint 3 (Train + Evaluate models)**
 - Training and test DistilBERT (FUS/non-FUS)
 - Training and test SVM, log regression, naive bayes (FUS/non-FUS)
 - Train and test another BERT (FUS/non-FUS)
 - Training and test multiclass BERT (indications)
 - ***Project Progress Report 1***

Updates 3/22/24

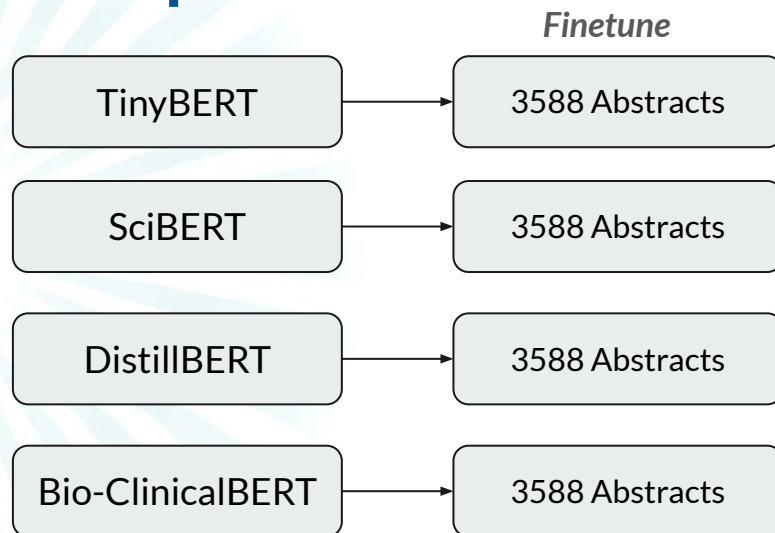
SPRINT 4 - Resolve Model Issues		02/29	03/27	42	28			
Compile supplementary data source (sean's data, fuf website data, zotero database) organize data sources	Rose	02/29	03/27	42	28	In Progress		
Clean data before tokenizing (get rid of background objective other filler words)	Rose	02/29	03/27	42	28	In Progress		
Create template of final report in overleaf	Reanna	02/29	03/27	42	28	Complete		
Start documenting process of model creation in overleaf final report	All	02/29	03/27	42	28	In Progress		
Start working on multi-class model	Skye	02/29	03/27	42	28	In Progress		
Make a workflow map → datasource → train-test split → model (epochs), other fine tuning steps...	Rose	02/29	03/27	42	28	In Progress		
Finish FUS/non-FUS classification (DistilBERT, old models, tinyBERT, medBERT, docBERT)	All	02/29	03/27	42	28	In Progress		

Updates 4/5/24

- **Updates on dataset - zotero_data.csv**
 - All articles in Zotero database (except veterinary)
 - Non-FUS - “ultrasound diagnostic” in PubMed article titles
- **Best model - SciBERT**
- **Tags instead of multiclass classification**
- **Next Sprint:**
 - Additional dataset from most recent Pubmed extraction
 - September 2023-March 2024
 - Example cases
 - Compiling/comparing models
 - Hyper-parameter fine-tuning
 - Model inferencing
 - Excel script
 - Finish SIEDS paper
 - Meet with data management team

```
{'eval_loss': 0.2066207230091095,  
'eval_accuracy': 0.947075208913649,  
'eval_f1': 0.9493333333333334,  
'eval_precision': 0.9035532994923858,  
'eval_recall': 1.0,  
'eval_runtime': 1.5817,  
'eval_samples_per_second': 226.966,  
'eval_steps_per_second': 14.541,  
'epoch': 1.0}
```


Recap



Training

2906 Abstracts
~80% of dataset

Hyper-params

323 Abstracts
~9% of dataset

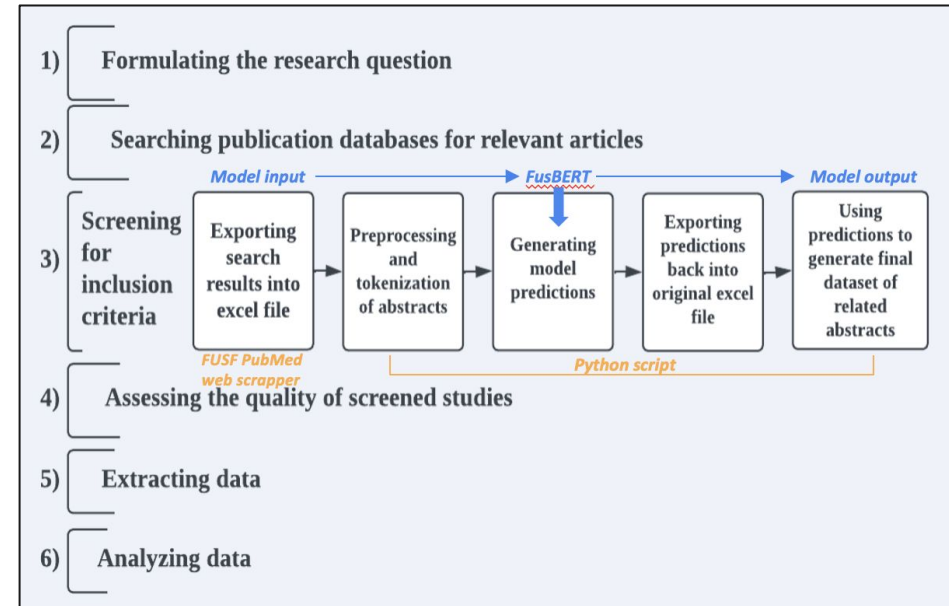
Testing

359 Abstracts
~10% of dataset

Recap

Model	Accuracy	Precision	Recall	F1
Naive Bayes	0.87	0.89	0.87	0.87
SVM	0.89	0.90	0.89	0.89
Logistic Regression	0.90	0.91	0.90	0.90
TinyBERT	0.88	0.82	0.98	0.89
SciBERT	0.89	0.89	0.99	0.90
DistilBERT	0.90	0.84	0.99	0.91
Bio-ClinicalBERT	0.91	0.85	0.99	0.91

TABLE IV: Performance comparison of different models



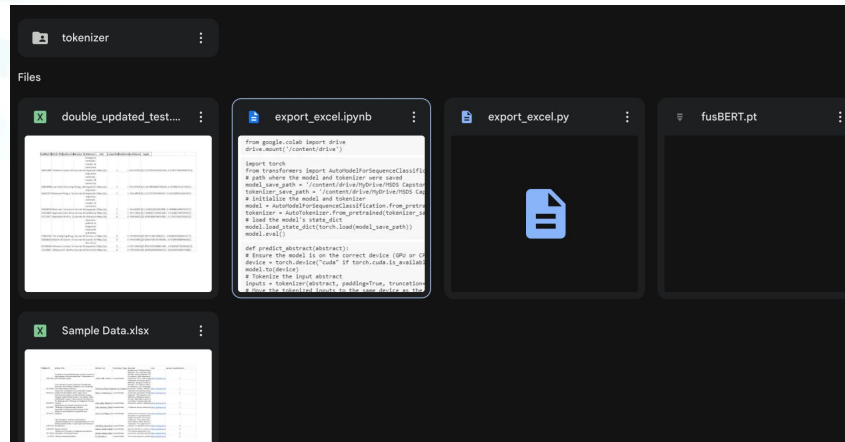
Recap

PubMed ID	Article Title	Author List	Publication Type	Abstract	Link	(actual classification)	Prediction	Confidence	Logits			
34921489	Treatment of super	Jacek Calik, Tomas	Journal Article	Background: Existing therapies Methods: Two volunteer subjects Results: During treatment, blood Conclusion: High intensity focused Keywords: HIFU; cherry angioma	https://pubmed.ncbi.nlm.nih.gov/34921489/	1	1	97.51520753	[[-1.67552316818881226, 1.994294285774231]]			
36444908	Low-Intensity Focused	Shenrong Zhong, Qianyun Cai, Longhe Z	Journal Article	Objectives: This study aims to Methods: Sprague-Dawley rats Results: LIFU reduced heart rate Conclusions: LIFU ameliorates	https://pubmed.ncbi.nlm.nih.gov/36444908/	1	1	97.46402502	[[-1.6663931608200073, 1.9825135469436646]]			
36541473	Autonomic modulation	Ning Ji, Yuanheng	Journal Article	Objective: Our previous study Keywords: autonomic modulation	https://pubmed.ncbi.nlm.nih.gov/36541473/	1	1	97.73917198	[[-1.7370295524597168, 2.0295417308807373]]			
35396078	American Association of	Sina Jasim, Kपाल	Journal Article	Objective: The objective of this Methods: A literature search Results: Minimally invasive therapy Conclusion: Surgery and radio	https://pubmed.ncbi.nlm.nih.gov/35396078/	1	1	97.89092541	[[-1.7765820026397705, 2.061021089553833]]			
35326687	Applications of Focused	John Panzone, Tina	Journal Article	Traditional cancer treatments	https://pubmed.ncbi.nlm.nih.gov/35326687/	1	1	97.96125293	[[-1.7842141389846802, 2.0880227088928223]]			
25753877	Application of ultra	W-X Fu, Q Wang, Y	Journal Article	Ultrasound is commonly used	https://pubmed.ncbi.nlm.nih.gov/25753877/	0	0	99.07332659	[[2.402134895324707, -2.2698824405670166]]			
37682185	The ambiguous "in	Jing Ning, Gang Zh	Journal Article	Rationale: Congenital anatom Patient concerns: A 61-year-old Diagnoses: The patient was di Interventions: After relevant p Outcomes: The patient recove Lessons: Congenital absence c	https://pubmed.ncbi.nlm.nih.gov/37682185/	0	0	99.40914512	[[2.568174362182617, -2.557249069213867]]			
22608626	Bipolar disorder	Daniel J Smith, Eli	Journal Article	Bipolar disorder is a serious di	https://pubmed.ncbi.nlm.nih.gov/22608626/	0	0	99.82409477	[[3.3475353717803955, -2.9937052726745605]]			
28748549	Ultrasound Curricula	Usman Tarique, Bi	Journal Article	The clinical applications of poi Keywords: curricula; education	https://pubmed.ncbi.nlm.nih.gov/28748549/	0	0	99.76135492	[[3.210164785385132, -2.8253657817840576]]			
3519082	Ultrasound instrum	R J Bartrum Jr	Journal Article	This article begins by reviewin	https://pubmed.ncbi.nlm.nih.gov/3519082/	0	0	99.5505929	[[2.7637088298797607, -2.636791467666626]]			

Demo

https://github.com/rteb8/MSDS_FUSCapstone23/tree/main

1. Access handover files
 - Google drive folder - [FUS handover](#)
 - Github repo - “handover” folder, access fusBERT.pt using [huggingface](#)
2. Modify + run `export_excel.ipynb`
 - We recommend using google colab to execute .ipynb file



Demo

https://github.com/rteb8/MSDS_FUSCapstone23/tree/main

1. Runtime to PyTorch 2.0.1, Hardware accelerator to T4-GPU
2. Install torch, transformers, pandas, openxyl packages
 - pip install torch, pip install transformers, pip install pandas, pip install openxyl
3. Make sure excel file is in folder/repository where the .ipynb file is
 - Use “*Sample Data.xlsx*” for demo
4. Edit .ipynb file to have the correct path, then run all cells in the file
 - model, tokenizer, input data, output file path

```
# path where the model and tokenizer were saved
```

```
model_save_path = '/content/drive/MyDrive/MSDS Capstone/FUS_handover/fusBERT.pt'
```

```
tokenizer_save_path = '/content/drive/MyDrive/MSDS Capstone/FUS_handover/tokenizer'
```

```
file_path = '/content/drive/MyDrive/MSDS Capstone/FUS_handover/Sample Data.xlsx' #update file path to relevant file  
df = pd.read_excel(file_path)
```

```
# Save to a new Excel file
```

```
output_file_path = '/content/drive/MyDrive/MSDS Capstone/FUS_handover/double_updated_test.xlsx'
```

```
df.to_excel(output_file_path, index=False)
```

April Data

Actual classification

```
0      419  
1       75  
Name: fus_related, dtype: int64
```

Model prediction

```
1      493  
0        1  
Name: Prediction, dtype: int64
```


Potential Causes

- BERT models have lower precision scores than recall scores → may need to achieve higher precision scores
 - More likely to predict a *false positive* than *false negative*
- Difference between non-FUS related articles BERT models used for training + monthly reviews
 - Initial dataset → 48 FUS related, 441 non-FUS
 - Final dataset → 1794 FUS related, 1794 non-FUS
- Articles on FUS changes over time (*concept drift*) → Feb-Aug 2023 + supplemental articles may be different than April 2024 articles

Recommendations

- Further work needs to be done before incorporating BERT models into literature review pipeline
 - BERT (language models) models are data intensive, initial dataset was too small and didn't have enough FUS-related articles to fine tune BERT models
 - Re-train BERT models → with only articles used for monthly literature review
 - Train models that are less data-hungry to perform classification
 - Assess difference between non-FUS articles in test dataset + April 2024 data
 - More guidance on labeling of “non-FUS” articles
 - Develop a feedback loop into the model pipeline which can incorporate human review of model output + learn from it
 - Address changes concept drift of FUS articles without needing to changing modeling approach

Future Work

- Multi-class classification of FUS manuscript indications
 - Modify PubMed web-scraper to pull in tags
- Explore other transformer architectures for text classification
- Develop feedback loop into modeling pipeline to address concept drift
- Incorporate “explainable AI” features into model prediction

Tags pulled in by Zotero

Info	Notes	Tags	Attachments	Related
16 tags				
<input type="radio"/> *Epilepsy/therapy				
<input type="radio"/> *Music Therapy				
<input type="radio"/> Acoustic Stimulation				
<input type="radio"/> Anticonvulsants				
<input type="radio"/> Auditory				
<input type="radio"/> Auditory Perception				
<input type="radio"/> Brain/diagnostic imaging				
<input type="radio"/> Electroencephalography				
<input type="radio"/> Epilepsy				
<input type="radio"/> Humans				
<input type="radio"/> Infrasound				
<input type="radio"/> Mozart				
<input type="radio"/> Music				
<input type="radio"/> Seizures				
<input type="radio"/> Sound				
<input type="radio"/> Ultrasound				

Questions?