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Project title:

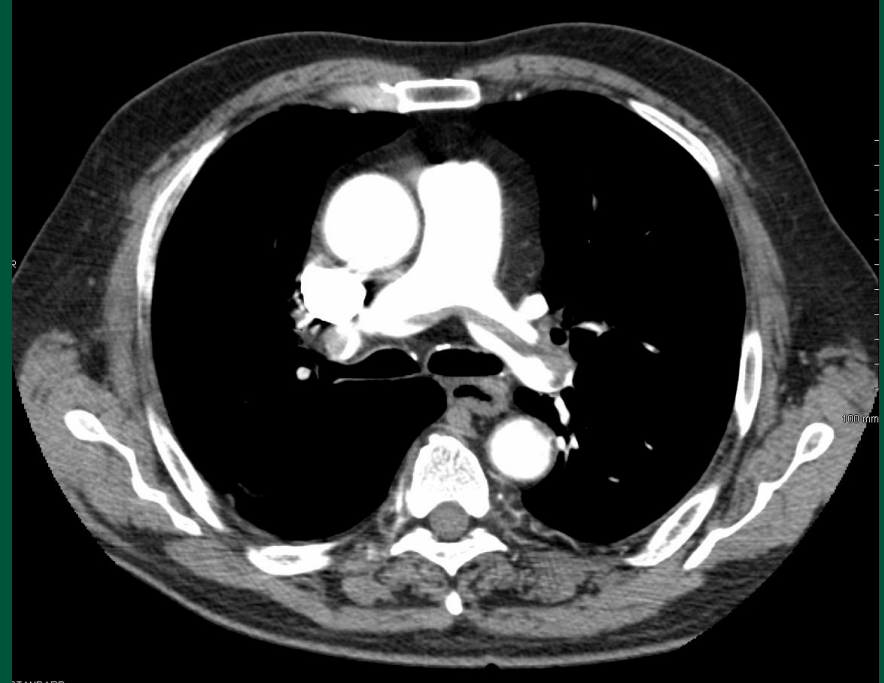
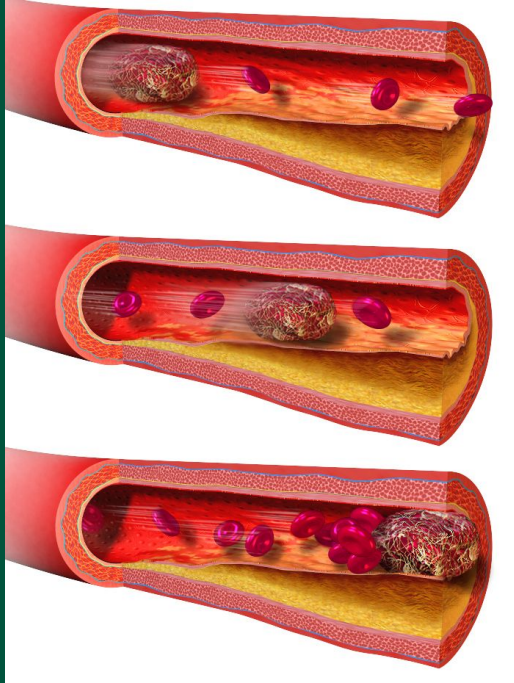
Detecting true cases of Pulmonary Embolism in MIMIC-III dataset.

Instructor: Andrew Nguyen

Student: Omid Khazaie

HS 619: NATURAL LANGUAGE PROCESSING
FALL 2018

What is Pulmonary Embolism (PE)?



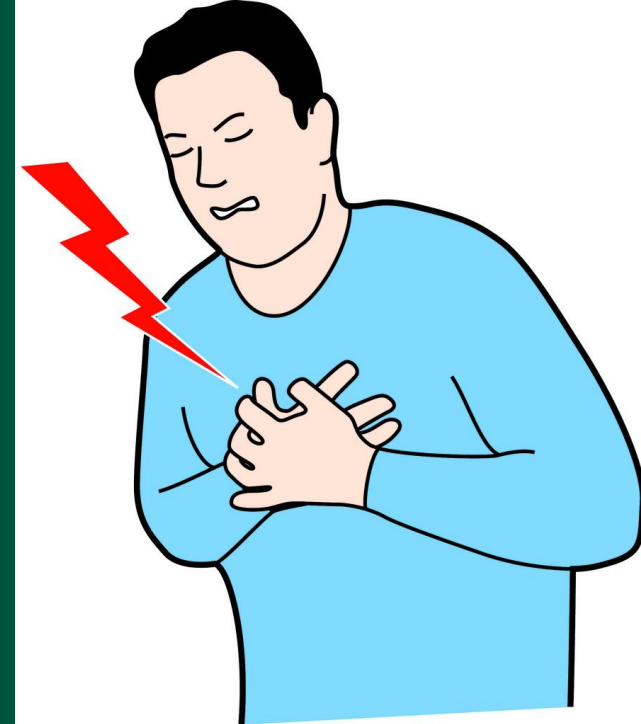
PE Symptoms

Common signs and symptoms include:

- Shortness of breath
- Chest pain
- Cough

Other signs and symptoms:

- Leg pain or swelling, or both, usually in the calf
- Clammy or discolored skin
- Fever
- Excessive sweating
- Rapid or irregular heartbeat
- Lightheadedness or dizziness [2]



Risk factors:

Medical history:

Family members history of venous blood clots or pulmonary embolism

Medical conditions and treatments:

- Heart disease
- Cancer
- Surgery

Prolonged immobility:

- Bed rest
- Long trips

Other risk factors:

- Smoking
- Being overweight
- Supplemental estrogen
- Pregnancy



Issues and Background

Recent study by patient safety experts at Johns Hopkins Medicine in Baltimore:

- 40,500 ICU adult patients a year die with an unknown medical condition
- Doctors receive about 7,000 pieces of information a day in this complex, distracting environment
- Misdiagnoses also occur frequently in emergency rooms, where doctors are scrambling to decide whether patients should be admitted to the hospital or sent home.
- Just five conditions account for more than one-third of all missed diagnoses in the ICU

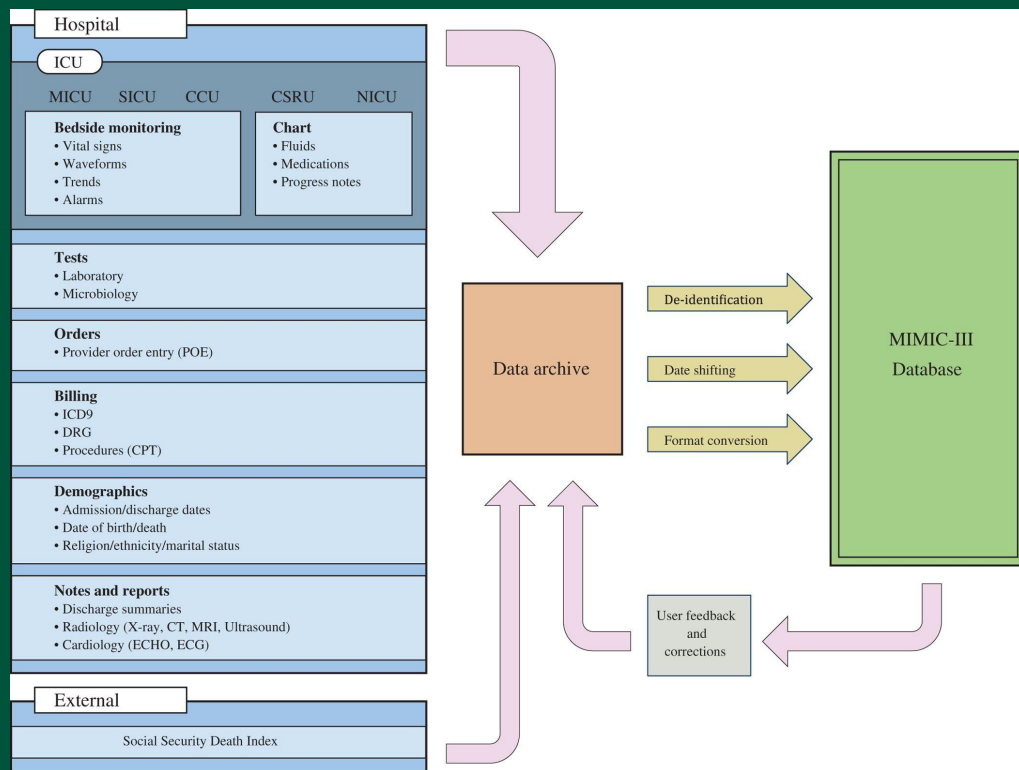
5 Most Commonly Misdiagnosed Conditions in the ICU:

1. Heart attack
2. Pulmonary embolism
3. Pneumonia
4. Aspergillosis
5. Abdominal bleeding



Dataset: MIMIC-III

- MIMIC is an openly available dataset developed by the MIT Lab for Computational Physiology
- Comprising de-identified health data associated with ~40,000 critical care patients.
- 26 tables
 - Track patient stays
 - Data collected in the critical care unit
 - Data collected in the hospital record system
 - Dictionaries



Data Extraction

noteevents table:

- Number of rows: 2,083,180
- TEXT is often large and contains many newline characters
- Some reports are available for both inpatient and outpatient stays
- If a patient is an outpatient, there will not be an HADM_ID associated with the note

Processing all the notes?

- Computationally expensive
- Not required for this project

How to extract data?

- Use ICD codes (billing code) to filter cases with PE or similar conditions
- Filter Discharge summary notes



Data Extraction

diagnoses_icd table:

- The ICD codes are generated for billing purposes at the end of the hospital stay.
- All ICD codes in MIMIC-III are ICD-9 based
- Number of rows: 651,047
- SEQ_NUM: provides the order in which the ICD diagnoses relate to the patient.

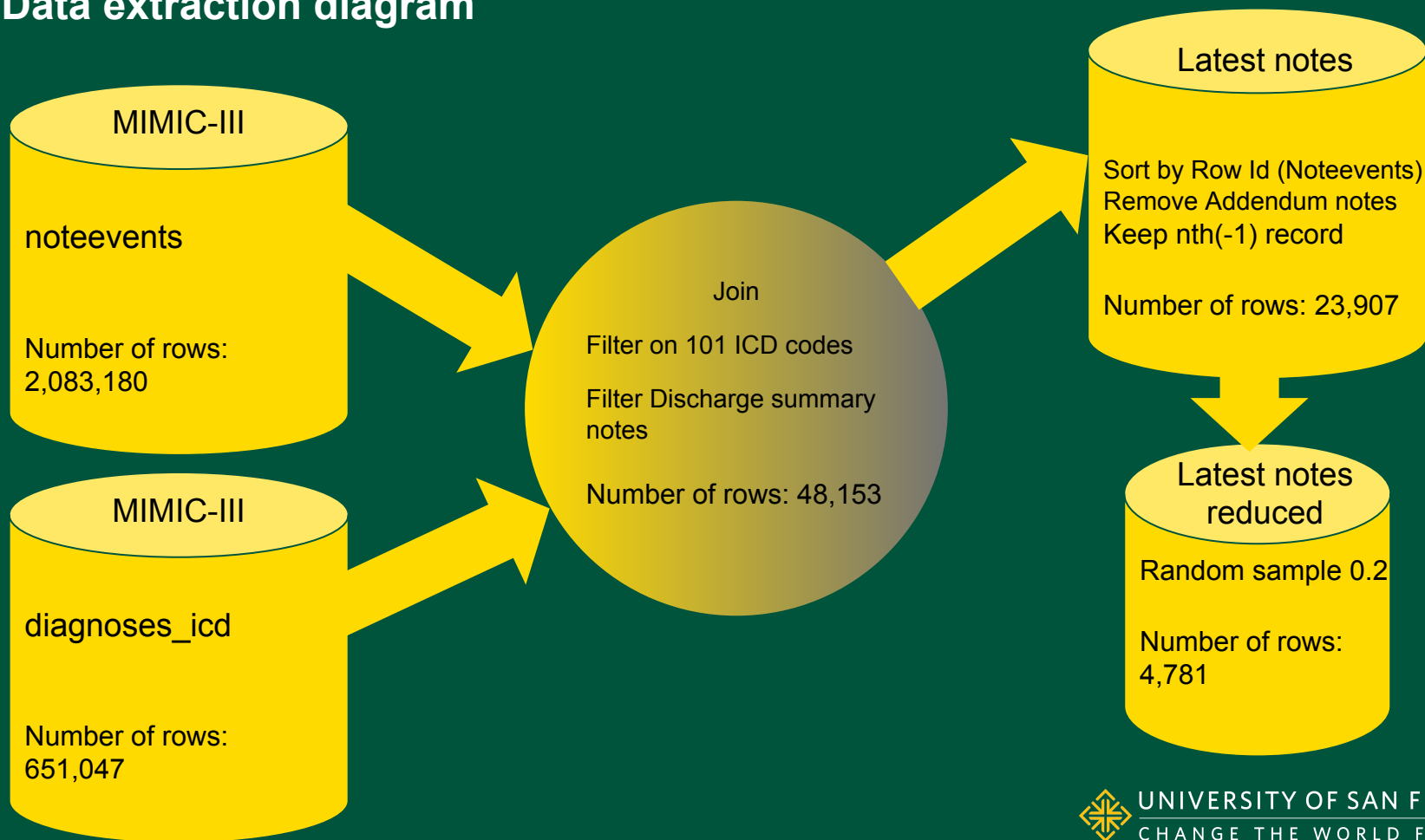
ICD codes used to extract data:

- Asthma, Pneumonia, Bronchitis, Heart attack
- Total of 101 ICD codes

Use the most recent note for each patient

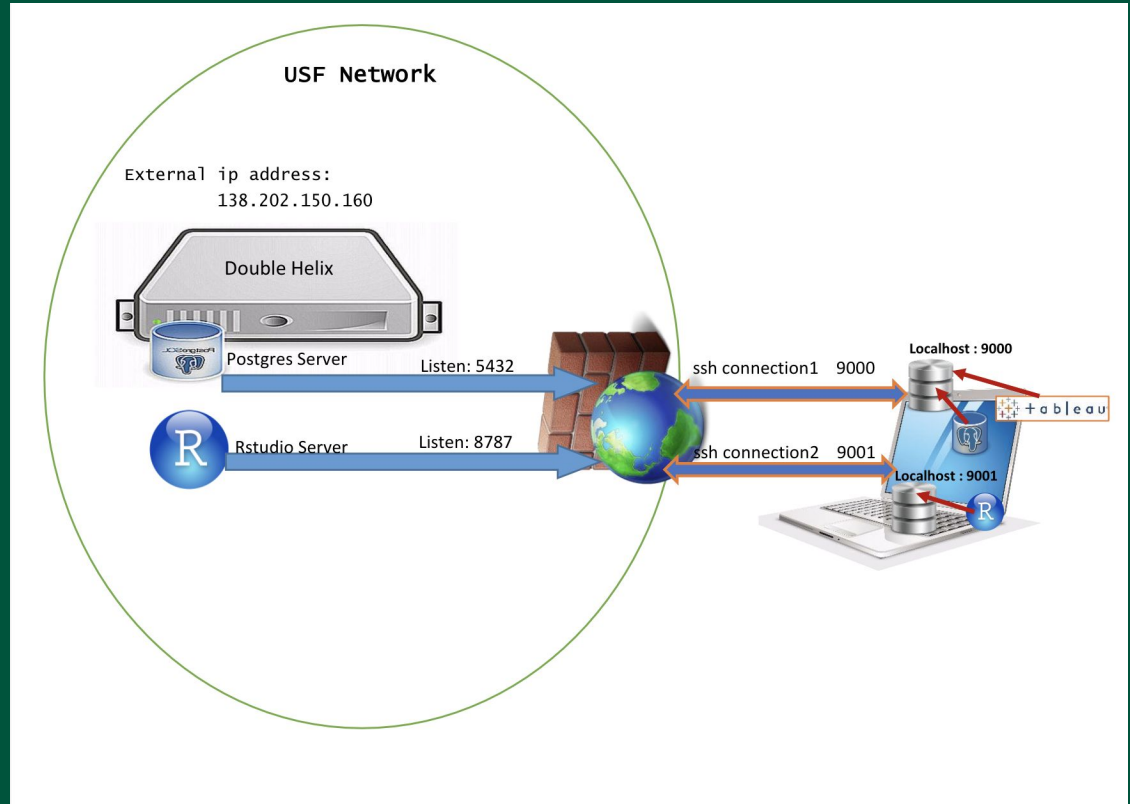


Data extraction diagram



Data extraction tools and techniques

- Postgres server
- pgAdmin
- Tableau
- IntelliJ
- R
- Spark
- Python



Remote ssh connection

Postgres:

```
ssh -L 9000:localhost:5432 mimicuser@138.202.150.160
```

```
psql -h localhost -p 9000 -U mimicuser -d mimic
```



The screenshot displays the IntelliJ IDEA interface. On the left, the 'Data Sources and Drivers' dialog box is open, showing the 'mimic@localhost' data source selected. The 'General' tab is active, displaying the following configuration:

- Name: mimic@localhost
- Comment:
- Host: localhost
- Port: 9000
- Database: mimic
- User: mimicuser
- Password: <hidden> (Remember password checked)
- URL: jdbc:postgresql://localhost:9000/mimic
- Driver: PostgreSQL

At the bottom of the dialog, it shows '59 objects' and 'Tx: Auto' with checkboxes for 'Read-only' and 'Auto sync'.

On the right, the 'databases' tree view is expanded, showing the 'mimic' database. The 'schemas' tree is also expanded, showing the 'mimiciii' schema. The 'tables' tree is expanded, showing a list of tables including 'admissions', 'callout', 'caregivers', 'charevents', 'charevents_1' through 'charevents_16', 'charevents_2' through 'charevents_9', 'cptevents', 'd.cpt', 'd.icd_diagnoses', 'd.icd_procedures', 'd.items', 'd.labitems', 'datetimeevents', 'diagnoses_icd', 'drgcodes', 'drgstays', 'inputevents_cv', 'inputevents_mv', 'labevents', 'microbiologyevents', and 'notevents'. The 'notevents' table is selected, showing its columns: 'row_id integer', 'subject_id integer', 'hadm_id integer', 'chartdate timestamp(0)', 'charttime timestamp(0)', 'storetime timestamp(0)', 'category varchar(50)', and 'description varchar(255)'.



Spark set-up and challenges

- Java 8
- Docker set-up
- Dependency hell

Project SDK:

This SDK is default for all project modules.

A module specific SDK can be configured for each of the modules as required.

 1.8 (java version "1.8.0_202") ▼

New...

Edit

Project language level:

This language level is default for all project modules.

A module specific language level can be configured for each of the modules as required.

8 - Lambdas, type annotations etc. ▼

Project compiler output:

This path is used to store all project compilation results.

A directory corresponding to each module is created under this path.

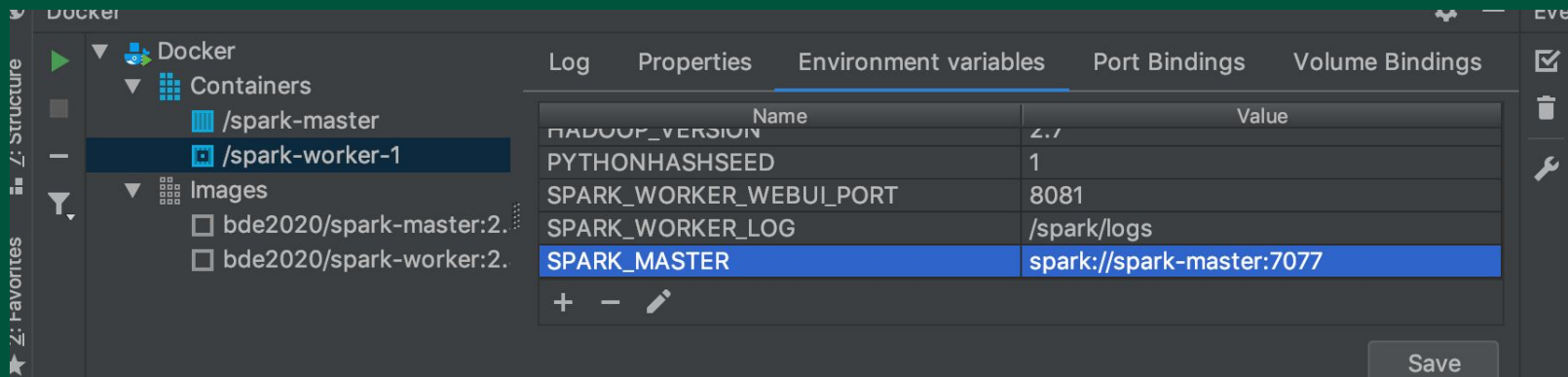
This directory contains two subdirectories: Production and Test for production code and test sources, respectively.

A module specific compiler output path can be configured for each of the modules as required.

/Users/user/619/uima-annotator/out



Docker set-up



The screenshot shows the Docker Desktop interface. On the left, the 'Containers' section is expanded, showing two containers: '/spark-master' and '/spark-worker-1'. The '/spark-worker-1' container is selected. The 'Environment variables' tab is active, displaying a table of environment variables. The 'SPARK_MASTER' variable is highlighted in blue.

Name	Value
HADOOP_VERSION	2.7
PYTHONHASHSEED	1
SPARK_WORKER_WEBUI_PORT	8081
SPARK_WORKER_LOG	/spark/logs
SPARK_MASTER	spark://spark-master:7077

Buttons for '+', '-', and a pencil icon are visible at the bottom of the table. A 'Save' button is located at the bottom right of the interface.

```
val spark = SparkSession
    .builder()
    .master(master = "spark://spark-master:7077")
    .appName(name = "Spark")
    .getOrCreate()
val conf = new SparkConf()
    .setMaster("spark://spark-master:7077")
    .setAppName("Spark")
    .setJars()
```



Dependency hell

cTAKES and Spark both were built on carrotsearch but Spark is using the older version of carrotsearch

```
libraryDependencies += commons.io % commons.io % 2.1.0
// https://mvnrepository.com/artifact/org.t3as/metamap-tagger
libraryDependencies += "org.t3as" % "metamap-tagger" % "1.3.4"
libraryDependencies += "au.com.bytecode" % "opencsv" % "2.4"
// https://mvnrepository.com/artifact/org.apache.spark/spark-core
libraryDependencies += "org.apache.spark" %% "spark-core" % "2.3.2" exclude("com.carrotsearch", "hppc")
libraryDependencies += "org.apache.spark" %% "spark-sql" % "2.3.2" exclude("com.carrotsearch", "hppc")
libraryDependencies += "org.apache.spark" %% "spark-streaming" % "2.3.2" % "provided" exclude("com.carrotsearch", "hppc")
libraryDependencies += "org.apache.spark" %% "spark-mllib" % "2.3.2" % "runtime" exclude("com.carrotsearch", "hppc")
libraryDependencies += "org.apache.spark" %% "spark-hive" % "2.3.2" % "provided" exclude("com.carrotsearch", "hppc")
```



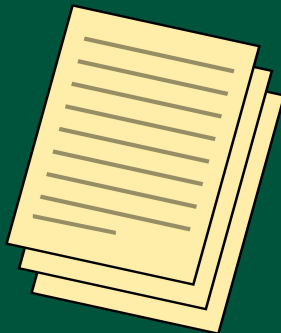
NLP pipeline

Input

cTAKES processing

Ontology processing

Output



Sentence detector
Tokenizer
Context dependent tokenizer
POS tagger
Chunker
UMLS dictionary look-up annotator
Dependency parser
Semantic role labeler

Cui resolver
Polarity resolver
Uncertainty resolver
Signsymptom resolver



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New NLP pipeline

Input

cTAKES processing

Ontology Concept processing

Output

Sentence detector

Tokenizer

Context dependent tokenizer

POS tagger

Chunker

UMLS dictionary look-up annotator

Dependency parser

Semantic role labeler

Cui resolver

Polarity resolver

Uncertainty resolver

Signsymptom resolver



OntologyConcept processing

```
>>>>>> print umls size: 2
>>>>>> printing cui: C0039239<<<<<< index: 0 <<<<<< high_index: 3
>>>>>> printing cui: C0039239<<<<<< index: 1 <<<<<< high_index: 3

>>>>>> print umls size: 12
>>>>>> printing cui: C0423772<<<<<< index: 0 <<<<<< high_index: 4
>>>>>> printing cui: C0016169<<<<<< index: 1 <<<<<< high_index: 4
>>>>>> printing cui: C0016169<<<<<< index: 2 <<<<<< high_index: 4
>>>>>> printing cui: C0423772<<<<<< index: 3 <<<<<< high_index: 4
>>>>>> printing cui: C0423772<<<<<< index: 4 <<<<<< high_index: 4
>>>>>> printing cui: C0016169<<<<<< index: 5 <<<<<< high_index: 4
>>>>>> printing cui: C0423772<<<<<< index: 6 <<<<<< high_index: 4
>>>>>> printing cui: C0016169<<<<<< index: 7 <<<<<< high_index: 4
>>>>>> printing cui: C0423772<<<<<< index: 8 <<<<<< high_index: 4
>>>>>> printing cui: C0423772<<<<<< index: 9 <<<<<< high_index: 4
>>>>>> printing cui: C0016169<<<<<< index: 10 <<<<<< high_index: 4
>>>>>> printing cui: C0423772<<<<<< index: 11 <<<<<< high_index: 4
```



OntologyConcept processing

```
val diseaseOrDisorders = JCasUtil.select(aJCAS, classOf[DiseaseDisorderMention])
val diseaseOrDisorders_array = diseaseOrDisorders.toArray(new Array[DiseaseDisorderMention](0))
for(d <- diseaseOrDisorders_array) {
    var umlsconcept = JCasUtil.select(d.getOntologyConceptArr(), classOf[UmlsConcept])
    var umlsconcept_array = umlsconcept.toArray(new Array[UmlsConcept](0))
    //println(">>>>>> print umls size: " + umlsconcept_array.size)
    var pos = 0

    for (con <- umlsconcept_array) {

        //println(">>>>>> printing cui: " + con.getCui + "<<<<<< index: " + pos + " <<<<<< ")
        if (con.getCui == "C0034065" | con.getCui == "C0919697" | con.getCui == "C2747923"
            | con.getCui == "C0157540" | con.getCui == "C1535887" | con.getCui == "C0151947"
            | con.getCui == "C0034074" | con.getCui == "C0520546" | con.getCui == "C2721578"
            | con.getCui == "C0151946" | con.getCui == "C4524050" | con.getCui == "C0392108"
            | con.getCui == "C1868769") {
            positive_case = true
            if (d.getPolarity == -1){
                polarity_case = true
            }
            if (d.getUncertainty == 1) {
                uncertainty_case = true
            }
            println(">>>>>> index: " + pos + " <<<<<< high_index: " + pos)
            println(">>>>>> CUI: " + con.getCui)
            println(">>>>>>>>>> POLARITY: " + d.getPolarity)
            println(">>>>>>>>>> Uncertainty: " + d.getUncertainty)
            println(">>>>>>>>>> Confidence: " + d.getConfidence + "\n\n")
        }
        pos += 1
    }
}
```



Features extracted

Out[77]:

	cui_exist	negation_exist	uncertainty_exist
0	0	0	0
1	0	0	0
2	1	0	0
3	0	0	0
4	0	0	0
5	0	0	0
6	0	0	0
7	0	0	0
8	0	0	0
9	1	0	0
10	0	0	0



What is my gold standard?



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Machine Learning - Gold Standard

The ICD codes are generated for billing purposes at the end of the hospital stay:

- 4150
- 41511
- 41512
- 41513
- 41519

	cui_exist	negation_exist	uncertainty_exist	hadm_id	seq_number	target
0	0	0	0	111544	16	0
1	1	0	0	166737	2	1
2	0	0	0	137804	15	0
3	1	1	0	162201	2	0
4	1	1	0	168769	6	0



ML Model- DNN

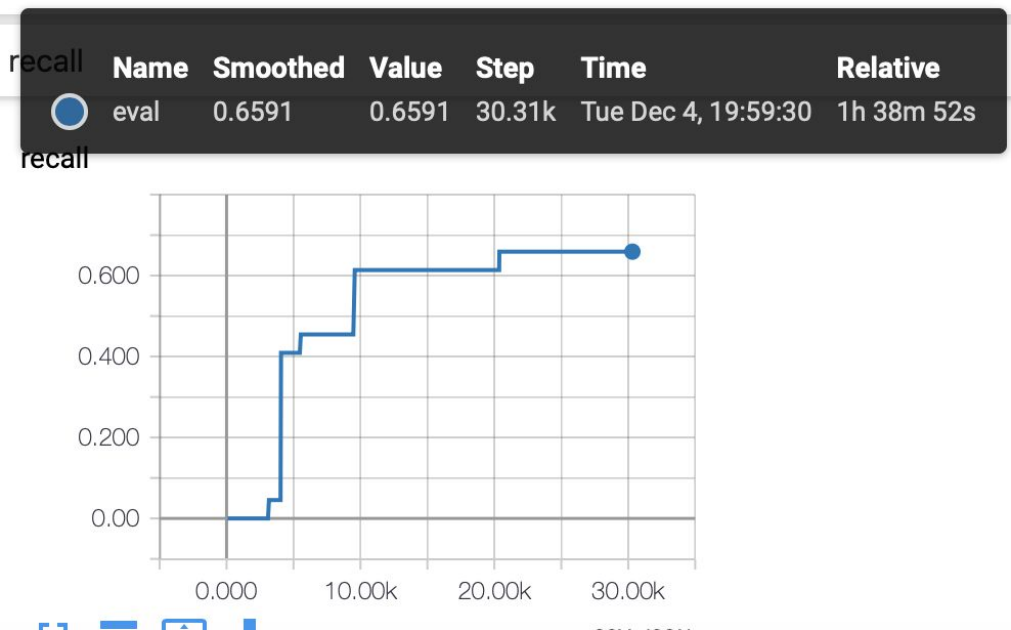
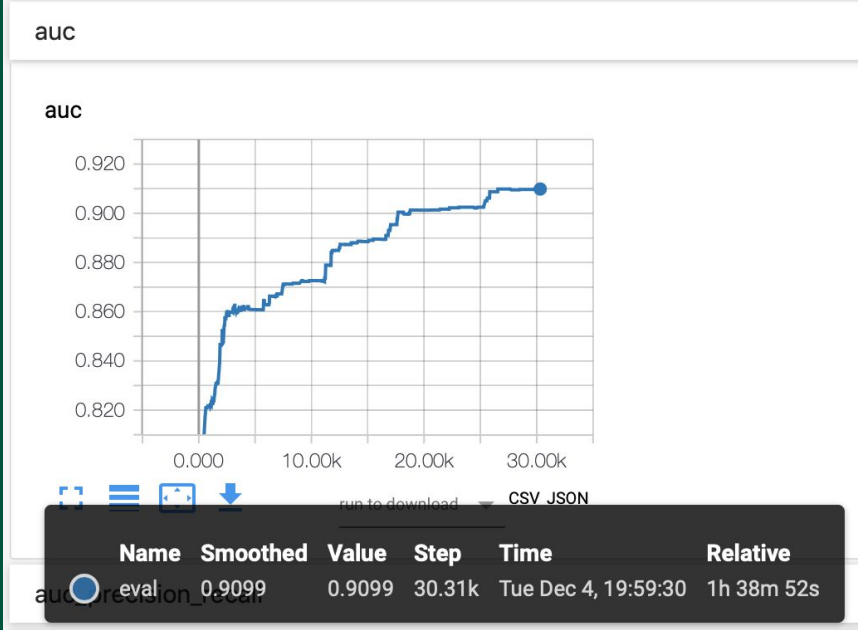
```
1 col_names = list(train_x.columns)
2 col_names.remove('hadm_id')
3 feature_columns = [tf.feature_column.numeric_column(key = key, shape=[1]) for key in col_names]
```

```
1 model_dir = os.path.join('.', 'model', 'dnn_classifier_threelayers_2')
```

```
1 model = tf.estimator.DNNClassifier(
2     feature_columns=feature_columns,
3     hidden_units=[32,16],
4     dropout=0.2,
5     model_dir=model_dir,
6     n_classes=2,
7     optimizer=tf.train.ProximalAdagradOptimizer(
8         learning_rate=0.001,
9         l1_regularization_strength=0.001))
```

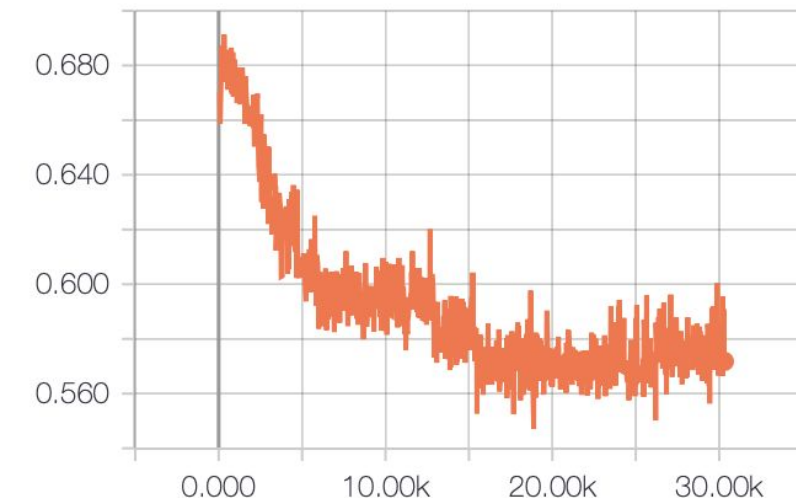


Results-AUC



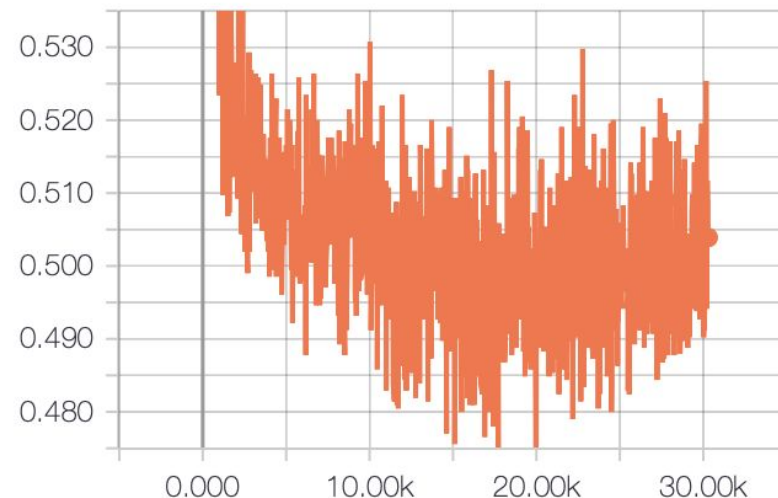
DNN-Layer activity

dnn/hiddenlayer_0/fraction_of_zero_values
tag: dnn/dnn/hiddenlayer_0/fraction_of_zero_values



run to download ▼ CSV JSON

dnn/hiddenlayer_1/fraction_of_zero_values
tag: dnn/dnn/hiddenlayer_1/fraction_of_zero_values



run to download ▼ CSV JSON



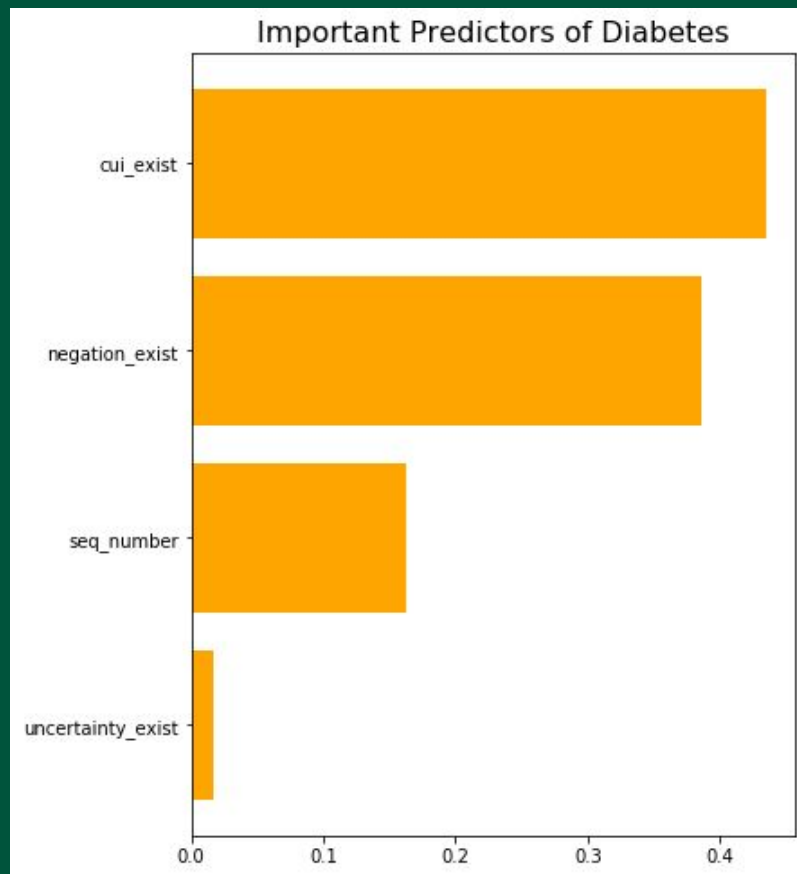
ML-Other models:

Imbalanced data: 0.07 have PE

Model	AUC with Negation and Uncertainty	AUC with Negation	AUC without Negation & Uncertainty
DNN	0.92	0.91	0.88
Gaussian Naive Bayes Classifier	0.82	0.82	0.80
Random Forest Classifier	0.79	0.80	0.78



RF Feature Importance



1. **Do Negation and Uncertainty improve our model?**
2. **Does training with gold standard make our result worse?**



	cui_exist	negation_exist	uncertainty_exist	hadm_id	icd_code	seq_number	target
72	0	0	0	185880	41519	1	1
354	0	0	0	129882	41519	2	1
363	0	0	0	141664	41519	2	1
370	0	0	0	147390	41519	2	1



Future work

Improve the results:

- Train the model with the gold standard (clinical notes reviewed by medical annotators)
- Extract more features from clinical notes:
 - Number of times cui found in the text
 - Number of time negation found in the text
 - Number of time uncertainty found in the text
 - Bag of cui's for sign and symptoms found in the text



Improve the model:

- Optimize the algorithm
- Try other biomedical semantic annotation tools like MetaMap



Optimize the algorithm $O(n^2) \rightarrow O(n)$

```
val diseaseOrDisorders = JCasUtil.select(aJCAS, classOf[DiseaseDisorderMention])
val diseaseOrDisorders_array = diseaseOrDisorders.toArray(new Array[DiseaseDisorderMention](0))
for(d <- diseaseOrDisorders_array) {
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  var umlsconcept_array = umlsconcept.toArray(new Array[UmlsConcept](0))
  //println(">>>>>> print umls size: " + umlsconcept_array.size)
  var pos = 0

  for (con <- umlsconcept_array) {
    //println(">>>>>> printing cui: " + con.getCui + "<<<<<< index: " + pos + " <<<<<< ")
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      | con.getCui == "C0157540" | con.getCui == "C1535887" | con.getCui == "C0151947"
      | con.getCui == "C0034074" | con.getCui == "C0520546" | con.getCui == "C2721578"
      | con.getCui == "C0151946" | con.getCui == "C4524050" | con.getCui == "C0392108"
      | con.getCui == "C1868769") {
      positive_case = true
      if (d.getPolarity == -1){
        polarity_case = true
      }
      if (d.getUncertainty == 1) {
        uncertainty_case = true
      }
      println(">>>>>> index: " + pos + " <<<<<< high_index: " + p)
      println(">>>>>> CUI: " + con.getCui)
      println(">>>>>>>>>> POLARITY: " + d.getPolarity)
      println(">>>>>>>>>> Uncertainty: " + d.getUncertainty)
      println(">>>>>>>>>> Confidence: " + d.getConfidence + "\n\n")
    }
    pos += 1
  }
}
```

sConcept])

mlsConcept](0)))

lem.getCui))



References:

1. <https://commons.wikimedia.org/wiki/File:SaddlePE.PNG>
2. <https://www.mayoclinic.org/diseases-conditions/pulmonary-embolism/symptoms-causes/syc-20354647>
3. <https://www.flickr.com/photos/easy-pics/9609168594>
4. <https://pixabay.com/en/anatomy-blood-vessel-red-156854/>



Thank you!



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