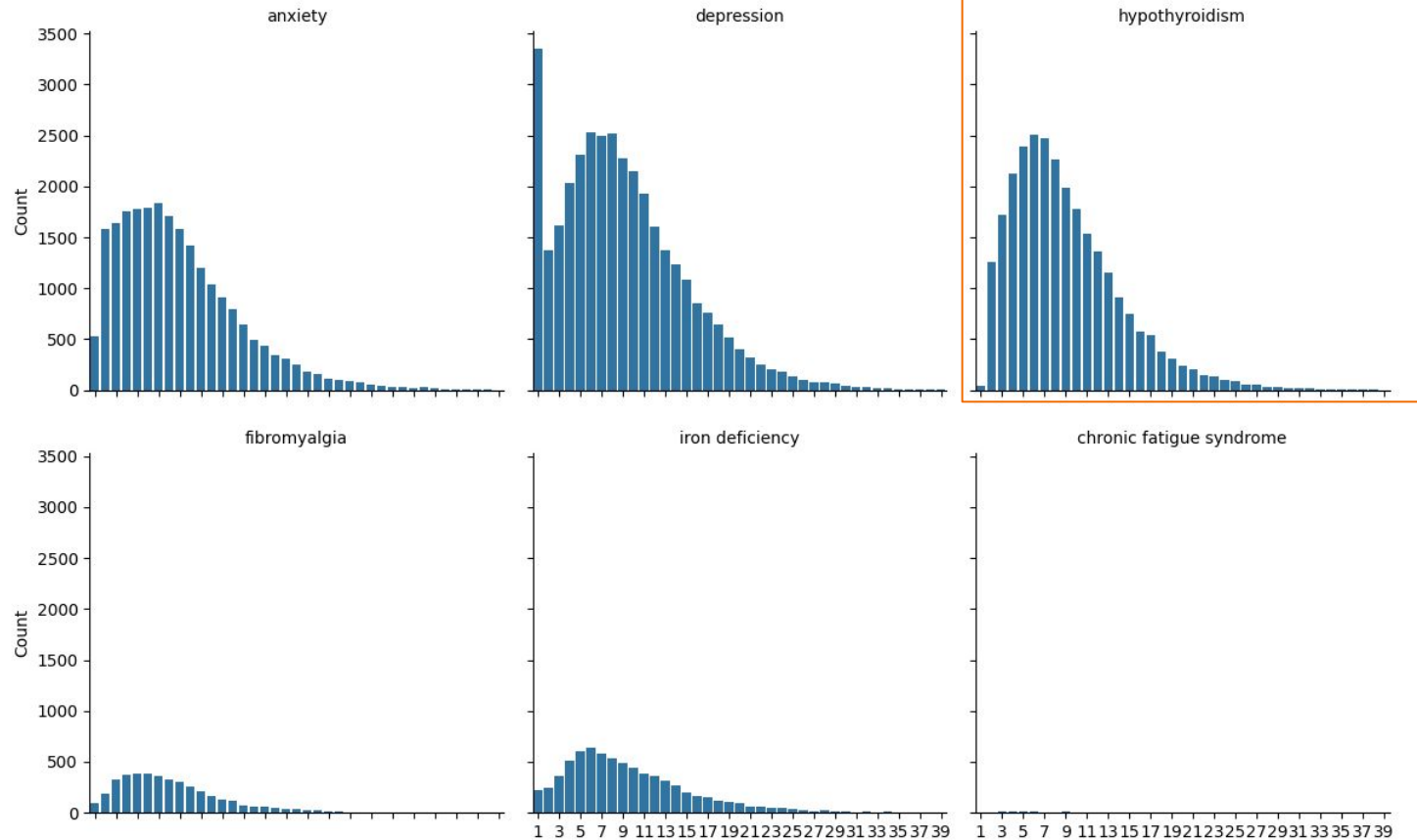

Diagnosing Hypothyroidism

— Hypothyroidism is one of the top 10
misdiagnosed diseases in America —

Thyroid Disease is number 6 on the list

<https://ml-law.net/medical-errors-misdiagnosis-negligence/10-most-commonly-medical-misdiagnosed-illnesses>

Number of Visits for each condition



Very few case of hypothyroidism are diagnosed with only one visit compared to some other disease especially depression.

Symptoms

- Fatigue
- Weight gain
- Depression
- Constipation
- A Sudden Spike in Cholesterol:

(High cholesterol is sometimes the only evidence of an underactive thyroid in an older person. Because this sign may stand alone, high cholesterol warrants a thyroid evaluation.)

<https://www.centrichealth.ie/health-wellness-blog/hypothyroidism-and-dementia-misdiagnosis-as-we-age/>

Hypothyroidism is often
mistaken for:

- Anxiety
 - Bipolar disorder-cyclothymia
 - Iron deficiency
 - Chronic fatigue syndrome
 - Fibromyalgia
 - Depression
-

MIMIC IV

Tables we choose

```
!cp drive/MyDrive/labevents.csv.gz ./  
!cp drive/MyDrive/omr.csv.gz ./  
!cp drive/MyDrive/diagnoses_icd.csv.gz ./  
!cp drive/MyDrive/d_labitems.csv.gz ./
```

```
#load necessary tables  
omr= pd.read_csv('omr.csv.gz',compression='gzip')  
diagnoses = pd.read_csv('diagnoses_icd.csv.gz',compression='gzip')  
lab_items = pd.read_csv('d_labitems.csv.gz',compression='gzip')
```

Omr: contains height and weight, bmi, blood pressure and eGFR. - Weight gain, high blood pressure and kidney function can all be indicators of hypothyroidism.

Diagnoses_icd: to see who is actually diagnosed with hypothyroidism.

Lab_items: see if we can use certain labs to indicate hypothyroidism.

```
#make list of codes want, add more to list if find any
codes_list = ['2448', '2449', '30000', '30001', '30002', '30009', '30113', '2801', '2808', '2809', '78071', '7291', '311']
#filter diagnoses table for codes on our list
diagnoses_filtered = diagnoses[diagnoses['icd_code'].isin(codes_list)]
#validate that have all of them
print(diagnoses_filtered['icd_code'].unique())
```

```
['30000' '311' '2449' '7291' '2809' '30001' '30009' '30002' '2448' '2808'
 '78071' '30113' '2801']
```

```
#make list of all subject id's selected to use for filtering other tables
subject_id_list = diagnoses_filtered['subject_id']
print(subject_id_list.shape)
```

```
(96280,)
```

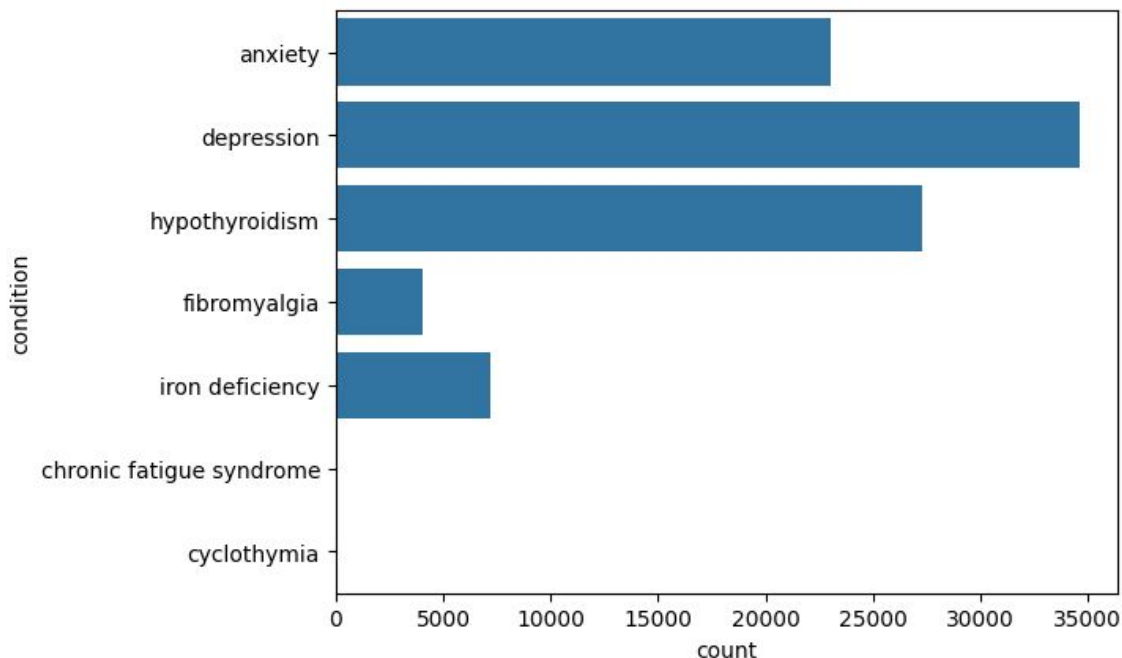
Filtering

Selecting only the ones for hypothyroidism and the few disease that it's often mistaken for.

How often does each condition appear?

```
#group by each condition
diagnoses_filtered.loc[diagnoses_filtered['icd_code'].isin(['2448', '2449']), 'condition'] = 'hypothyroidism'
diagnoses_filtered.loc[diagnoses_filtered['icd_code'].isin(['30000', '30001', '30002', '30009']), 'condition'] = 'anxiety'
diagnoses_filtered.loc[diagnoses_filtered['icd_code'].isin(['30113']), 'condition'] = 'cyclothymia'
diagnoses_filtered.loc[diagnoses_filtered['icd_code'].isin(['2801', '2808', '2809']), 'condition'] = 'iron deficiency'
diagnoses_filtered.loc[diagnoses_filtered['icd_code'].isin(['78071']), 'condition'] = 'chronic fatigue syndrome'
diagnoses_filtered.loc[diagnoses_filtered['icd_code'].isin(['7291']), 'condition'] = 'fibromyalgia'
diagnoses_filtered.loc[diagnoses_filtered['icd_code'].isin(['311']), 'condition'] = 'depression'
print(diagnoses_filtered['condition'].unique())
print(diagnoses_filtered.head())
```

```
sns.countplot(diagnoses_filtered, y= 'condition')
plt.show()
```



**Labs item table is only a text description for each item id.
We need the labevents table.**

```
#check what lab items table  
looks like
```

```
print(lab_items.head())
```

```
print(len(lab_items))
```

```
print(lab_items['category'].unique()  
())
```

```
print(lab_items['label'].nunique()  
())
```

```
print(lab_items['label'].unique()  
())
```

```
print(lab_items['fluid'].unique()  
())
```

	itemid	label	fluid	category
0	50801	Alveolar-arterial Gradient	Blood	Blood Gas
1	50802	Base Excess	Blood	Blood Gas
2	50803	Calculated Bicarbonate, Whole Blood	Blood	Blood Gas
3	50804	Calculated Total CO2	Blood	Blood Gas
4	50805	Carboxyhemoglobin	Blood	Blood Gas

```
1622  
['Blood Gas' 'Chemistry' 'Hematology']  
1170  
['Alveolar-arterial Gradient' 'Base Excess'  
'Calculated Bicarbonate, Whole Blood' ... 'Anti-la' 'HIV FINAL'  
'HIV Screen']  
['Blood' 'Other Body Fluid' 'Ascites' 'Joint Fluid' 'Pleural' 'Stool'  
'Urine' 'Bone Marrow' 'Cerebrospinal Fluid' 'Fluid' 'I' 'Q']
```


Loading lab events table.

The lab events table is very large and the notebook kept on crashing as there was not enough ram available. We experimented using different chunk sizes and 10 chunks worked.

When we further reduced it to 5 it crashed again.

```
# Create a TextFileReader object by reading the CSV file in chunks
#10 chunks worked when reduced to 5 it crashed
labs_reader = pd.read_csv('labs_events.csv.gz', compression='gzip', chunksize=11820000)
```

it working on first chunk now. In end might save each one to separate csv files and combine the filtered version so it doesn't crash.

```
#name chunk
chunk1 = next(labs_reader)
print(chunk1.head())
```

	labevent_id	subject_id	hadm_id	specimen_id	itemid	order_provider_id	\
0	1	10000032	NaN	45421181	51237	P28Z0X	
1	2	10000032	NaN	45421181	51274	P28Z0X	
2	3	10000032	NaN	52958335	50853	P28Z0X	
3	4	10000032	NaN	52958335	50861	P28Z0X	
4	5	10000032	NaN	52958335	50862	P28Z0X	

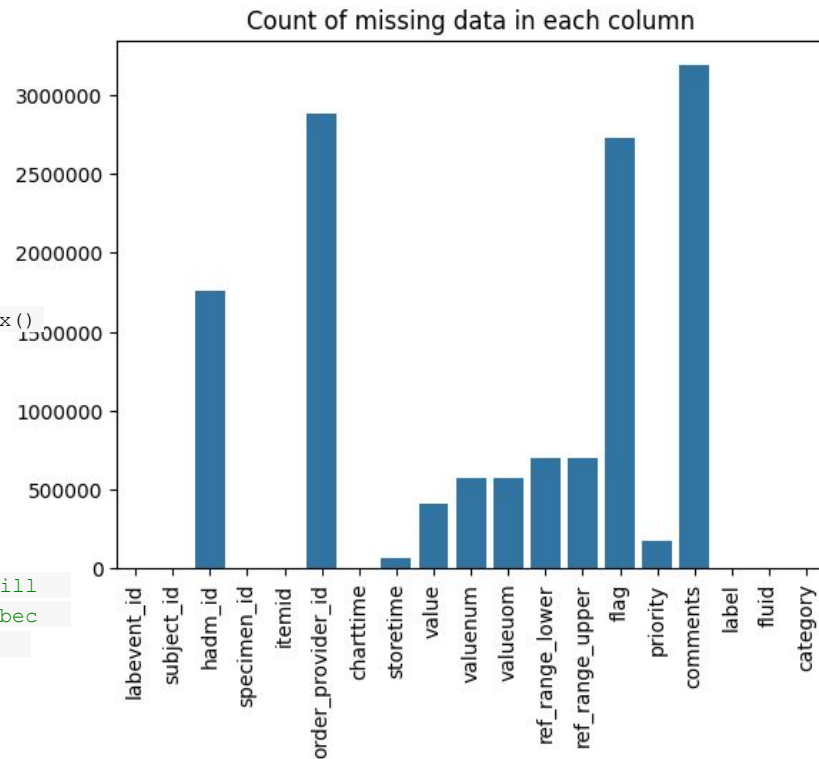
	charttime	storetime	value	valuenum	valueuom	\
0	2180-03-23 11:51:00	2180-03-23 15:15:00	1.4	1.4	NaN	
1	2180-03-23 11:51:00	2180-03-23 15:15:00	___	15.1	sec	
2	2180-03-23 11:51:00	2180-03-25 11:06:00	___	15.0	ng/mL	
3	2180-03-23 11:51:00	2180-03-23 16:40:00	102	102.0	IU/L	
4	2180-03-23 11:51:00	2180-03-23 16:40:00	3.3	3.3	g/dL	

	ref_range_lower	ref_range_upper	flag	priority	\
0	0.9	1.1	abnormal	ROUTINE	
1	9.4	12.5	abnormal	ROUTINE	
2	30.0	60.0	abnormal	ROUTINE	
3	0.0	40.0	abnormal	ROUTINE	
4	3.5	5.2	abnormal	ROUTINE	

	comments
0	NaN
1	VERIFIED.
2	NEW ASSAY IN USE ___: DETECTS D2 AND D3 25-OH ...
3	NaN
4	NaN

Null values

```
#filter first_chunk_merg for patients from diagnosed table
labs_chunk1_merg_filter =
first_chunk_merg[first_chunk_merg[ 'subject_id' ].isin(subject_id_list)]
#check if worked and is reduced
print(labs_chunk1_merg_filter.shape)
print(first_chunk_merg.shape)
print(labs_chunk1_merg_filter.shape)
print(labs_chunk1_merg_filter.isna().sum())
data_na = pd.DataFrame(labs_chunk1_merg_filter.isna().sum()).reset_index()
#plot missing data
sns.barplot(data=data_na, x= 'index',y=0)
plt.ticklabel_format(style= 'plain', axis='y')
plt.xticks(rotation= 90)
plt.title('Count of missing data in each column' )
plt.ylabel( "")
plt.xlabel( "")
plt.show()
#remove hadm_id and order_provider_id and comments too many null values will
also drop timings for now.also remove ref range lower and upper for now. bec
seems like if not in range it's flagged as abnormal go off that value for
now.
labs_chunk1_merg_filter
=labs_chunk1_merg_filter.drop([ 'hadm_id', 'order_provider_id',
                                'comments',
                                'charttime', 'storetime', 'ref_range_lower', 'ref_range_upper' ],axis = 1)
```



Value column also has text which is hard to run models on. The value num is only the ones with integers. Values that are not within normal range are flagged. For now will just work with that and use all values that flagged as abnormal.

```
labs_chunk1_merg_filter  
=labs_chunk1_merg_filter.drop(['valuenum',  
'valueuom','label' 'fluid','category',  
'labevent_id', 'specimen_id','value'  
,axis = 1)  
print(labs_chunk1_merg_filter.head())
```

	subject_id	itemid	flag	priority
2644	10000826	50955	normal	STAT
2645	10000826	51237	abnormal	STAT
2646	10000826	51274	abnormal	STAT
2647	10000826	51275	normal	STAT
2648	10000826	51146	normal	STAT

Filter to only use the ones that are flagged as abnormal. Once only have abnormal ones can remove that column and changed priority column to 1 if it was marked as stat.

```
#filter to only the ones which are abnormal and try to see which tests  
with abnormal results can predict hypothyroidism  
labs_chunk1_merg_filter =  
labs_chunk1_merg_filter[labs_chunk1_merg_filter['flag'] == 'abnormal']  
print(labs_chunk1_merg_filter.shape)  
#drop flag column since only have abnormal now  
labs_chunk1_merg_filter =labs_chunk1_merg_filter.drop(['flag'],axis = 1)  
#turn priority column stat into 1  
# Define the mapping dictionary  
priority_mapping = {'STAT': 1, 'ROUTINE': 0}  
  
# Replace values using the mapping dictionary  
labs_chunk1_merg_filter['priority'] =  
labs_chunk1_merg_filter['priority'].replace(priority_mapping)  
  
# Print unique values to verify the transformation  
print(labs_chunk1_merg_filter['priority'].unique())  
  
#change name to priority stat  
labs_chunk1_merg_filter =  
labs_chunk1_merg_filter.rename(columns={'priority': 'priority_stat'})  
  
# Print the updated DataFrame to verify the change  
print(labs_chunk1_merg_filter.head())
```

Filter to only include top 20 tests that have abnormal results and that person was diagnosed with hypothyroidism

'INR(PT)''PT''Hematocrit''Hemoglobin''MCH''MCHC'
'MCV''Neutrophils''Platelet Count''RDW''Red Blood
Cells''White Blood Cells''Bicarbonate''Chloride'
'Creatinine''Glucose''Urea Nitrogen'
'Calcium, Total''Phosphate''RDW-SD'

Get top 20

```
# Initialize an empty list to store the mode for 'hypothyroidism'
for each round
    mode_list = []
    hypo_mode_list = []

# Loop for 20 rounds
for i in range(1, 21):

    # Group by 'condition' and find the mode of 'itemid' for each
    condition
    stats =
merged_labdg.groupby('condition')['itemid'].agg(mode=lambda x:
x.mode().iloc[0])

# Append the mode for 'hypothyroidism' to the list
stats.reset_index(inplace=True)
hypothyroidism_mode = stats.iloc[5,1]
mode_list.append(stats)
hypo_mode_list.append(hypothyroidism_mode)

# Remove rows where 'itemid' is equal to the mode for
'hypothyroidism'
merged_labdg = merged_labdg[merged_labdg['itemid'] !=
hypothyroidism_mode]

# Print the DataFrame
print(hypo_mode_list)
```

Split testing and training

Had to split subject id first to ensure each person stays in same group.

```
from sklearn.model_selection import train_test_split
# Split id's into training and testing sets for when
split data
X_trainid, X_testid = train_test_split(subject_id_list,
test_size=0.25, random_state=42)
# Print the shapes of the training and testing sets
print(X_trainid.shape)
print(X_testid.shape)
#split merged labdg according to id to ensure that each
person stays in one group
X_train_labs =
merged_labdg[merged_labdg['subject_id'].isin(X_trainid)]
X_test_labs =
merged_labdg[~merged_labdg['subject_id'].isin(X_trainid)]
print(X_train_labs.shape)
print(X_test_labs.shape)
#divide each into x and y. Y being condition
y_train_omr = X_train_omr['condition']
y_test_omr = X_test_omr['condition']
y_train_labs = X_train_labs['condition']
y_test_labs = X_test_labs['condition']
```

Models

Created a function to run the models.

Included logistic regression, gradient boosting, and decision tree

```
def evaluate_models(X_train, X_test, y_train, y_test):  
    #import models  
    #logistic regression  
    from sklearn.linear_model import LogisticRegression  
    #gradient boosting  
    from sklearn.ensemble import GradientBoostingRegressor  
    # decision tree  
    from sklearn.tree import DecisionTreeRegressor  
    from sklearn.metrics import average_precision_score # try different models  
  
    #list of models  
  
    models=[LogisticRegression(),GradientBoostingRegressor(),DecisionTreeRegressor()  
    ]  
    #empty list to store avg precision  
    avg_precisions = []  
  
    #run models  
    for model in models:  
  
        #fit the model  
        model.fit(X_train,y_train labs)  
        y_pred = model.predict(X_test)  
        # Calculate the average precision of the model  
        avg_precision = average_precision_score(y_test, y_pred)  
  
        # Append the average precision to the list  
        avg_precisions.append(avg_precision)  
  
    # Create a dataframe of the average precision of each model  
    df_avg_precisions = pd.DataFrame({ 'Model': models, 'Average Precision' :  
    avg_precisions})  
  
    # Print the dataframe  
    print(df_avg_precisions)
```

Model Results

	Model	Average Precision
0	LogisticRegression()	0.361179
1	([DecisionTreeRegressor(criterion='friedman_ms...	0.385699
2	DecisionTreeRegressor()	0.384679

Gradient boosting model worked best at .385 avg precision

- Next steps:(need more time for these)
- Would want to optimize model to get it to perform better
- Also had omr data that didn't manage to merge yet or run model on. But we had started getting the data ready.
- Would use omr to figure out it have weight gain as well and use that metric.

Omr table

Has few values for blood pressure as well as height and weight consolidated all that.

```
#consolidate all blood pressure values
omr_filtered.loc[omr_filtered['result_name'].isin(['Blood Pressure Sitting', 'Blood Pressure Standing (1 min)', 'Blood Pressure Lying', 'Blood Pressure Standing (3 mins)', 'Blood Pressure Standing'])],
['result_name'] = 'Blood Pressure'
print(omr_filtered['result_name'].unique())
```

