Imports and utility functions

```
In [4]:
import numpy as np
import re
from pandas import DataFrame
from sqlalchemy import create engine
from matplotlib import pyplot as plt
import matplotlib as mpl
from sqlalchemy import create engine
from matplotlib import pyplot as plt
import cav
import denials config # contains sensitive data (git ignor)
from sklears.feature_extraction.text import TfidfVectorizer, CountVectorizer, TfidfTransformer
from sklearn.svm import LinearSVC
from sklearn import cross validation
from sklears.linear model import LogisticRegression
from sklears.metrics import precision score
from sklears.metrics import recall score
from sklearn import linear model, decomposition, datasets
from sklearn import svm
from sklears.metrics import roc_curve, auc
from sklearn.maive bayes import GaussianNB
from sklearn.ensemble import RandomForestClassifier
from sklearn.ensemble import SVC
plt.style.use('ggplot')
from matplotlib import rcParama
rcParams['font.family'] = 'sans-serif'
rcParams['font.serif'] - ['Times New Roman']
tmatplotlib inline
pd.set_option('display.max_colwidth', -1)
In [ ]:
def train_model(X, y, classifier):
   model = classifier.fit(X train, y train)
    precision = precision score(v test, model.predict(X test))
    recall = recall_score(y_test, model.predict(X_test))
    print("Training accuracy is {0}".format(model.score(X_train, y_train)))
print("Teating accuracy is {0}".format(model.score(X_test, y_test)))
print("Precision is {0}".format(precision))
print("Recall is {0}".format(recall))
    return model, X test, V test, X train, V train
In [20]:
# Simple function to prettify chart axes
def simpleaxis(ax):
   ax.spines['top'].set_visible(False)
ax.spines['right'].set_visible(False)
     ax.get_xaxis().tick_bottom()
   ax.get yaxis().tick left()
```

Data downloading

First, we pull down all of the patents that have ever been brought before the PTAB and join with patent text.

```
# Connecting to the db
host db = "cal-patent-lab.chhaitskv8dr.us-west-2.rds.amazonaws.com"
username - denials config.team_username
password - denials config.team password
db = "teamrocket"
engine - create engine("mwsgl://():()#()/()".format(
 username, password, host db, db))
connection = engine.connect()
q = "SELECT case id, patent id, invalidated, denied, filing date, decision date FROM ptab cases"
ptab patents = pd.read sql(q, engine)
connection.close()
In [19]:
# Next, we want to grab the claims text associated with each patent and insert it into the dataframe
host db = "rosencrantz.berkelev.edu"
username - denials config.fung username
password - denials config.fung password
db = "uspto"
engine - create_engine("mysql://{}:{}@{}/{}".format(
   username, password, host db, db))
connection = engine.connect()
patent id = int(ptab patents.patent id[:1])
found = list()
for idx, patent_id in enumerate(ptab_patents.patent_id):
   q = "SELECT text from uspto.claim where patent_id = '{}'".format(patent_id)
    claims = pd.read_sql(q, engine)
    claims concat = claims.text.str.cat()
    ptab patents.loc(ptab patents.patent id == '{}'.format(patent id), 'claim text' | = claims concat
    found.append(not claims.empty)
    if idx % 100 -- 0:
     print("Processing patent {}".format(idx))
In (1061:
# Pickling file for later use -- uncomment this line to overwrite the existing pickle
# ptab patents.to pickle('ptab patents.pkl')
Overall data description histogram
In [32]:
ptab_patents = pd.read_pickle('./pickles/ptab_patents.pkl')
num_denied = ptab_patents[ptab_patents.denied == 1].count()[0]
2052
Data cleaning
In (151:
ptab patents = pd.read pickle('./pickles/ptab patents.pkl')
# First get rid of patents where we cannot find the claim text
mask = (ptab patents.claim text.str.len() > 1)
df = ptab_patents.loc[mask]
null denied counts = df.denied.notnull().walue counts()
print("Out of {0} total cases, {1} have null for their denied status".format(sum(null denied counts), null
_denied_counts[1]))
Out of 4273 total cases, 2490 have null for their denied status
```

```
In [16]:
```

null_denied_counts

True 2490

False 1783 Name: denied, dtype: int64

In [18]:

dup_patent_counts = df.duplicated(subset='patent_id', keep='first').value_counts()
print("{0} patents show up more than once".format(dup_patent_counts[1]))

1656 patents show up more than once

Inferring denied status when possible

In (21):

First we want to represent the null denial status with an actual value - we'll use -1 for this (0 means not denied.) means denied)

df.denied = df.denied.fillna(value=-1)

df.invalidated = df.invalidated.fillna(value=-1)

/Users/davidjwiner/anaconda3/envs/py27/lib/python2.7/site-packages/pandas/core/generic.py:2701: SettingNit hCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc(row_indexer,col_indexer) = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy solf-nemp = value

Out[21]: -1.0 1783 0.0 1527

1.0 963 Name: invalidated, dtype: int64

In [22]:

stacked = df.groupby(by=['denied', 'invalidated']).count()

unstacked = stacked.unstack()

unstacked['case_id']

Out[22]:

| | invalidated | -1.0 | 0.0 | 1.0 |
|--|-------------|--------|-------|-------|
| | denied | | | |
| | -1.0 | 1783.0 | NaN | NaN |
| | 0.0 | NaN | 140.0 | 963.0 |
| | | | | |

As we'd expect, almost all of the cases that were not derived had a ruling. There are no cases that were both invalidated and derived (which makes sense). We'd lose a significant amount of data by calling all cases that are derived. "We can get around this by counting all of the cases that have an invalidation decision as not denied."

In [23]:

df['filing_to_decision'] = df.decision_date - df.filing_date

//users/davidjwiner/anaconda3/envs/py27/lib/python2.7/site-packages/ipykernel/_main_.py:1: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc(row indexer.col indexer) = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy if name == '_nain_':

```
In [24]:

yap in, days - df.filing to_decision[df.filing_to_decision.monthl()] / sp.timedelta64(), 'D')

ar - plt.seploc([1])

yap in, days, his (line20)

plt.title('Days elapsed from filing to_decision date for 'NFTAB cases, 2011-2016')

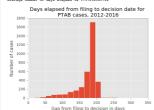
plt.title('Oays elapsed from filing to_decision in days')

plt.title('Days elapsed from filing to_decision in days')

plt.title('Days elapsed from filing to_decision in days')

Average number of_days elapsed is (0). *Gromat(spg_in_days.mesn())

Average number of_days elapsed is (0). *Gromat(spg_in_days.mesn())
```



So it looks like, for all cases from more than 200 days prior to when this case was filed, if there is no decision we can count them as "denied" for the purposes of our algorithm.

```
In [25])

# Eliminate any missing within the acceptable filling-to-decision gap

#f.decision data = pd.to_descimn(eff.decision_data)

#f.filling_data = pd.to_descimn(eff.filling_data)

#f.filling_data = pd.to_descimn(eff.filling_data)

#f.filling_data = pd.to_descimn(eff.filling_data)

#f.filling_data = ram_data = pd.filmedsia('200 daya')

#f.filling_data = ram_data = pd.filmedsia('200 daya')

#f.wooling_data = rdf((eff.decision = -1) = (eff.filling_data).commx()(0)

#f.wooling_data = rdf((eff.decision data = ram_filling_data)).commx()(0)

#f.wooling_data = ram_filling_data = ram_filling_data).commx()(0)
```

```
In [25]:
df -df([df.denied == 1) | (df.denied == 0) | ((df.denied == -1) & (df.filing_date < min_filing_date)))
df.denied[df.denied == -1] = 1</pre>
```

```
In (33):

df.deside.walow_counts()

Out[3]):

1.0 2541

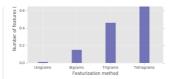
Marse deside.dysper_int64
```

That increased the amount that we have to work with from -2500 to -3600 - a pretty substantial increase!

Feature engineering

Removing stop words and applying stemmer

```
In (361:
X = df.clain_text.as_matrix()
y = df.denied.as matrix()
from nltk.corpus import stopwords
from nltk.stem.snowball import SnowballStemmer
ps = SnowballStemmer('english')
stop = set(stopwords.words('english'))
X_mod = []
for idx, claim in enumerate(X):
       text = ' '.join([ps.stem(word) for word in claim.split() if word not in stop1)
       text = ' '.join([word for word in claim.split() if word not in stop])
    X_mod.append(text)
Plotting number of features by featurization method
In (21):
tfidf_unigram = TfidfVectorizer(ngram_range=(1, 1))
tfidf_bigram = TfidfVectorizer(ngram_range=(1, 2))
tfidf_trigram = TfidfVectorizer(ngram_range=(1, 3))
tfidf tetragram = TfidfVectorizer(ngram range=(1, 4))
X unigram - tfidf unigram.fit transform(X mod)
X_bigram = tfidf_bigram.fit_transform(X)
X_trigram = tfidf_trigram.fit_transform(X)
X_tetragram = tfidf_tetragram.fit_transform(X)
In [ ]:
num_features = [feature_matrix.shape(1] for feature_matrix in [X_unigram, X_bigram, X_trigram, X_tetragram]
In [42]:
# Plotting number of features
from matplotlib import pyplot as plt
pos = list(range(len(num features)))
width = 0.3
fig, ax = plt.subplots()
fig.tight layout()
plt.bar([p + width for p is pos],
            num_features,
            width,
            alpha=0.5,
            color='darkblue',
            label-'Training accuracy')
ax.set vlabel('Number of features (millions)')
ax.set_xlabel('Featurization method')
ax.set_title('Number of features by featurization method', fontsize-12)
plt.ticklabel_format(style='sci', axis='y', scilimits=(0,0))
ax.set xticks([p + 1.5 * width for p in pos])
ax.set_xticklabels(['Unigrams', 'Bigrams', 'Trigrams', 'Tetragrams'])
(<matplotlib.text.Text at 0x12e5bf3d0>.
<matplotlib.text.Text at 0x12e4b86d0>,
<matplotlib.text.Text at 0x12e620090>,
<matplotlib.text.Text at 0x12e620810>1
                  Number of features by featurization method
    1.0 -1e6
   0.8
```



Singular value decomposition: effects of reducing the number of features

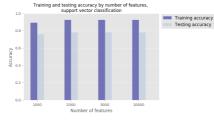
```
from sklearn import linear_model, decomposition, datasets
from sklearn import sym
from matplotlib import pyplot as plt
from sklearn import cross validation
plt.style.use('qqplot')
num svd features = [1000, 2000, 5000, 10000]
X train, X test, v train, v test - cross validation.train test splits
    X_unigram, y, test_size=0.2, random_state=20)
def plot_accuracies(train_acc, test_acc, figure, classifier name):
    pos = list(range(len(train acc)))
    width = 0.2
    plt.figure(figure)
    fig, ax = plt.subplots()
    fig.tight layout()
    plt.bar([p + width for p im pos],
            train_acc,
            width,
            alpha=0.5,
            color='darkblue'.
            label='Training accuracy')
    plt.bar([p + 2 * width for p in pos],
            test_acc,
            width.
            alpha=0.5.
            color='lightsteelblue',
            label = 'Testing accuracy')
    ax.set_ylabel('Accuracy')
ax.set_xlabel('Number_of_features')
    ax.set title('Training and testing accuracy by number of features, \n(0)'.format(classifier name), fon
   ax.set xticks([p + 2 * width for p in pos])
    ax.set_xticklabels(num svd features)
    plt.legend(bbox_to_anchor=(1, 1.02), loc='upper left', ncol=1)
    plt.show()
svc = svm.LinearSVC(C=10)
clfs = [('support vector classification', svc)]
for idx, (name, clf) in enumerate(clfs):
   training_accuracies = []
    testing_accuracies - []
    for n_components in num_svd_features:
        print("Working on componenents {0}".format(n components))
        svd = decomposition.TruncatedSVD(n_components-n_components)
        X train transformed = svd.transform(X train)
        X_test_transformed = svd.transform(X_test)
```

```
clf.fit(X_train_transformed, y_train)
training_accuracies.append(clf.score(X_train_transformed, y_train))
     testing_accuracies.append(clf.score(X_test_transformed, y_test))
plot_accuracies(training_accuracies, testing accuracies, idx, name)
```

Testing accuracy

plot accuracies(training accuracies, testing accuracies, idx, name)

<matplotlib.figure.Figure at 0x12ccc4c50>



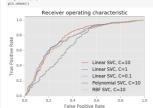
Primary classification results

```
In f 1:
from sklears.metrics import roc_curve, auc
from sklearn.naive bayes import GaussianNB
from sklearn.ensemble import RandomForestClassifier
from sklearn.svm import SVC
import matplotlib
import cay
# SVC classification algorithms
SVC_clfs = [("Linear SVC, C=10", SVC(C=10, kernel='linear')),
         ("Linear SVC, C=1" , SVC(C=1, kernel='linear')),
         ("Linear SVC, C=0.1", SVC(C=0.1, kernel='linear')),
("Polynomial SVC, C=10", SVC(C=10, kernel='poly')),
         ("RBF SVC, C=10", SVC(C=10, kernel='rbf'))]
random_forest_clfs = [("Random forest, 10", RandomForestClassifier(max_features=10, max_depth=10)),
         ("Random forest, 20", RandomForestClassifier(max features=10, max depth=20)),
("RandomS forest, 30", RandomForestClassifier(max features=10, max depth=30)),
         ("Random forest, 60", RandomForestClassifier(max features=10, max depth=60))]
def get_results(clfs, filename):
     for (name, classifier) in clfs:
         print(name)
         model, X test, y test, X train, y train = train model(X mod, y, classifier)
with open(filename, 'w') as csvfile:
             writer = csv.writer(csvfile)
              writer.writerow(['Training accuracy', 'Testing accuracy'])
              writer.writerow([model.score(X_train, y_train), model.score(X_test, y_test)])
qet results(random forest clfs, "random forest classifier accuracies.csv")
```

ROC curves

```
In [1]

The Control of the Control o
```



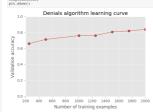
plt.ylabel('True Positive Rate')
plt.title('Receiver operating characteristic')
plt.legend(loc-"lower right")

simpleaxis(ax)

Learning curve

In []:

```
In [180]:
f, ax = plt.subplots()
plt.plot(n training examples[1:], testing accuracies[1:], 'o-')
plt.vlim(0, 1)
plt.xlabel("Number of training examples")
plt.ylabel("Validation accuracy")
plt.title("Denials algorithm learning curve")
simpleaxis(ax)
```



Repeated patents

```
In [208]:
```

top_patents = df.groupby(by='patent_id').count().sort(columns='case_id', ascending=False).case_id /Users/davidjwiner/anaconda3/envs/py27/lib/python2.7/site-packages/ipykernel/ main .py:1: FutureWarning: sort(columns=....) is deprecated, use sort_values(by=.....) if __name__ -- '__main__':

```
In [275]:
```

```
freqs - top_patents
patents - top_patents.index
sub frems = frems(frems > 1)
sub_patents = sub_freqs.index
```

for patent in sub patents: num denials.append(sum(df[df.patent id == patent].denied))

Tn (2791-

num denials - []

consensus = [num_denials[i] -- sub_freqs[i] for i in range(len(num_denials))] num patents without consensus = float(sum(consensus)/float(len(num denials)))

print("For {0} percent of patents that are duplicated across cases we have conflicting decisions".format(n un patents without consensus))

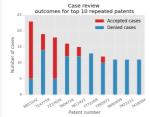
For 0.544827586207 percent of patents that are duplicated across cases we have conflicting decisions

```
In [220]:
```

```
freqs = top_patents[:10]
patents = top_patents[:10].index
num denials = []
```

for patent in patents: num denials.append(sum(df[df.patent id == patent].denied))

In [255]:



Getting most/least predictive words from SVM classifier

```
coeffs = nobal.coef 

In [40]:

sorted_labels = [7 for (x, y) is sarted(zip(coeffs.todense(), names))]

manages = coeffs.nonare(||1]

out = []
```

```
for idx in nonzeros:
   out.append((coeffs[0, idx], names[idx]))
sorted labels = [v for (x, v) in sorted(out)]
sorted coeffs = sorted(coeffs.todense().tolist()[0])
In (421:
x_vals = sorted_coeffs[-10:-1] + sorted_coeffs[0:9]
least_likely_denied = sorted_labels[-10:-1]
most_likely_denied = sorted_labels[0:9]
simpleaxis(ax)
from pylab import "
pos = arange(len(x_vals)) # the bar centers on the y axis
f, axarr = plt.subplots(2, sharex=True)
plt.suptitle('Relative influences on probability of denial \m (top and bottom word stems from TFIDF)')
for ax in axarr:
    simpleaxis(ax)
x 1 = sorted coeffs[-10:-1]
x_1_pos = arange(len(x_1))
a - axarri01
a.barh(x_1_pos, x_1)
a.set yticklabels(
# Customize minor tick labels
a.set_yticks(x_l_pos + 0.5, minor=True)
a.set_yticklabels(least_likely_denied, minor=True)
a.tick_params(axis='both', which='both',length=0)
a - axarrill
x_2 = sorted_coeffs[0:9]
x_2_pos = arange(len(x_2))
a.barh(x_2_pos, x_2)
a.set_yticklabels('
a.set_yticks(x_2_pos + 0.5, minor=True)
a.set yticklabels(most likely denied, minor-True)
a.tick_params(axis='y', which='both',length=0)
plt.xlabel('Regression coefficient \n (positive means word stem improves likelihood of denial, \nnegative
means stem hurts likelihood)')
plt.show()
                           Relative influences on probability of denial
                             (top and bottom word stems from TFIDF)
           name
          collect
```

