

Notebook

April 20, 2020

0.0.1 Question 1c

Discuss one thing you notice that is different between the two emails that might relate to the identification of spam.

The email not indicated as spam is nicely worded and ends with a thank you to the receiver of the email, while the email indicated as spam is worded like a shady commercial that is appealing to certain people.

0.0.2 Question 3a

Create a bar chart like the one above comparing the proportion of spam and ham emails containing certain words. Choose a set of words that are different from the ones above, but also have different proportions for the two classes. Make sure to only consider emails from `train`.

```
In [ ]: train = train.reset_index(drop=True) # We must do this in order to preserve the ordering of emails

words = ['body', 'business', 'html', 'money', 'offer', 'please']
filtered = words_in_texts(words, train['email'])
form = np.matrix(filtered)
df = pd.DataFrame(form)
df = df.rename(columns = {0: 'body', 1: 'business', 2: 'html', 3: 'money', 4: 'offer', 5: 'please'})
df['type'] = train['spam']
df = df.melt('type')
df['type'] = df['type'].map({0: 'ham', 1: 'spam'})

sns.barplot(x = 'variable', y = 'value', hue = 'type', data = df, ci = None)
```


0.0.3 Question 3b

Create a *class conditional density plot* like the one above (using `sns.distplot`), comparing the distribution of the length of spam emails to the distribution of the length of ham emails in the training set. Set the x-axis limit from 0 to 50000.

```
In [ ]: df = pd.DataFrame({
        'length': train['email'].apply(len), 'type': train['spam']
    })
df = df.melt('type')
df['type'] = df['type'].map({0: 'ham', 1: 'spam'})

plt.xlim(0, 50000)
x = df[df['type'] == 'ham']
y = df[df['type'] == 'spam']
one = sns.distplot(x['value'], label = 'ham', hist = False)
two = sns.distplot(y['value'], label = 'spam', hist = False)
plt.legend()
```