Multi-Label Text Classification Using Bidirectional LSTM (RNN)

In [2]: !pip install kaggle
 from google.colab import files
 files.upload()

Requirement already satisfied: kaggle in /usr/local/lib/python3.6/dist-packages (1.5.1.1)

Requirement already satisfied: urllib3<1.23.0,>=1.15 in /usr/local/lib/python3.6/dist-packages (from kaggle) (1.22)

Requirement already satisfied: six>=1.10 in /usr/local/lib/python3.6/dist-packa ges (from kaggle) (1.11.0)

Requirement already satisfied: certifi in /usr/local/lib/python3.6/dist-package s (from kaggle) (2018.11.29)

Requirement already satisfied: python-dateutil in /usr/local/lib/python3.6/dist -packages (from kaggle) (2.5.3)

Requirement already satisfied: requests in /usr/local/lib/python3.6/dist-packag es (from kaggle) (2.18.4)

Requirement already satisfied: tqdm in /usr/local/lib/python3.6/dist-packages (from kaggle) (4.28.1)

Requirement already satisfied: python-slugify in /usr/local/lib/python3.6/dist-packages (from kaggle) (2.0.1)

Requirement already satisfied: idna<2.7,>=2.5 in /usr/local/lib/python3.6/dist-packages (from requests->kaggle) (2.6)

Requirement already satisfied: chardet<3.1.0,>=3.0.2 in /usr/local/lib/python3. 6/dist-packages (from requests->kaggle) (3.0.4)

Requirement already satisfied: Unidecode>=0.04.16 in /usr/local/lib/python3.6/d ist-packages (from python-slugify->kaggle) (1.0.23)

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```
Saving sample_submission.csv to sample_submission.csv
Saving test_data.csv to test_data.csv
Saving train_data.csv to train_data.csv
Saving train_label.csv to train_label.csv
```

[1] Loading data

```
In [0]: import numpy as np
import pandas as pd

train = pd.read_csv("train_data.csv")
  label = pd.read_csv("train_label.csv")
  test = pd.read_csv("test_data.csv")
  sample_submission = pd.read_csv("sample_submission.csv")
```

[2] Data Preparation

```
In [5]: from sklearn import preprocessing
        lb = preprocessing.LabelBinarizer()
        lb.fit(label['label'])
Out[5]: LabelBinarizer(neg label=0, pos label=1, sparse output=False)
In [6]: lb.classes
Out[6]: array(['Assembly Required', 'Color', 'Commercial / Residential',
                'ENERGY STAR Certified', 'Features', 'Finish',
               'Flooring Product Type', 'Hardware Included', 'Included',
                'Indoor/Outdoor', 'Package Quantity', 'Shape',
                'Tools Product Type', 'Voltage (volts)', 'Wattage (watts)'],
              dtype='<U24')
In [0]: | a = lb.transform(label['label'])
In [0]: classes = pd.DataFrame(a, columns=['Assembly Required', 'Color', 'Commercial / Re
                'ENERGY STAR Certified', 'Features', 'Finish',
                'Flooring Product Type', 'Hardware Included', 'Included',
                'Indoor/Outdoor', 'Package Quantity', 'Shape',
                'Tools Product Type', 'Voltage (volts)', 'Wattage (watts)'])
```

```
result = pd.concat([label, classes], axis=1)
In [10]:
           result.head()
Out[10]:
                                                  Commercial
                                                              ENERGY
                                                                                         Flooring
                                                                                                  Hardware
                                Assembly
                                           Color
                   id
                                                                        Features Finish
                                                                                         Product
                          label
                                                                 STAR
                                 Required
                                                                                                   Included
                                                  Residential
                                                              Certified
                                                                                            Type
              100003
                                        0
                                               0
                                                           0
                                                                     0
                                                                               0
                                                                                      0
                                                                                               0
                                                                                                         (
                         Shape
                        Voltage
               100004
                                        0
                                               0
                                                           0
                                                                     0
                                                                               0
                                                                                      0
                                                                                               0
                                                                                                         (
                         (volts)
                        Wattage
              100004
                                        0
                                               0
                                                           0
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                                                                               0
                                                                                      0
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                         (watts)
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              100006
                                        0
                                               0
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                                                                     0
                                                                               0
                                                                                      0
                                                                                               0
                                                                                                         (
                         (watts)
                       ENERGY
                                                                     1
                                                                               0
                                                                                               0
                                        0
                                               0
                                                           0
                                                                                      0
                                                                                                         (
              100007
                          STAR
                        Certified
          res = result.drop(['label'], axis=1)
 In [0]:
 In [0]:
           df = pd.merge(train, res, on=['id'])
```

[3] Text in each category

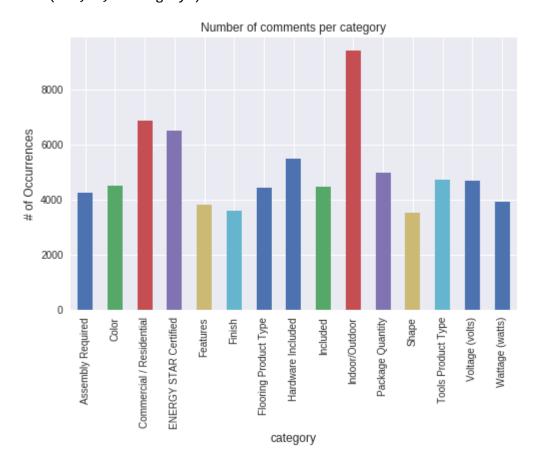
```
In [86]: df_label = df.drop(['id', 'text'], axis=1)
    counts = []
    categories = list(df_label.columns.values)
    for i in categories:
        counts.append((i, df_label[i].sum()))
    df_stats = pd.DataFrame(counts, columns=['category', 'number_of_comments'])
    df_stats
```

Out[86]:

	category	number_of_comments
0	Assembly Required	4266
1	Color	4505
2	Commercial / Residential	6866
3	ENERGY STAR Certified	6509
4	Features	3821
5	Finish	3602
6	Flooring Product Type	4422
7	Hardware Included	5471
8	Included	4465
9	Indoor/Outdoor	9420
10	Package Quantity	4995
11	Shape	3516
12	Tools Product Type	4708
13	Voltage (volts)	4688
14	Wattage (watts)	3916

```
In [87]: df_stats.plot(x='category', y='number_of_comments', kind='bar', legend=False, gri
    plt.title("Number of comments per category")
    plt.ylabel('# of Occurrences', fontsize=12)
    plt.xlabel('category', fontsize=12)
```

Out[87]: Text(0.5, 0, 'category')



[4] Text Preprocessing

```
In [0]: # https://stackoverflow.com/a/47091490/4084039
import re

def decontracted(phrase):
    # specific
    phrase = re.sub(r"won't", "will not", phrase)
    phrase = re.sub(r"can\'t", "can not", phrase)

# general
    phrase = re.sub(r"\'re", " are", phrase)
    phrase = re.sub(r"\'re", " is", phrase)
    phrase = re.sub(r"\'d", " would", phrase)
    phrase = re.sub(r"\'l", " will", phrase)
    phrase = re.sub(r"\'t", " not", phrase)
    phrase = re.sub(r"\'t", " not", phrase)
    phrase = re.sub(r"\'r", " not", phrase)
    phrase = re.sub(r"\'r", " and", phrase)
    return phrase
```

```
In [0]: # https://gist.github.com/sebleier/554280
    # we are removing the words from the stop words list: 'no', 'nor', 'not'
    # <br /><br /> ==> after the above steps, we are getting "br br"
    # we are including them into stop words list
    # instead of <br /> if we have <br/> these tags would have revmoved in the 1st stom stop words exet(['br', 'the', 'i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'our 'you'll", 'you'd", 'yours', 'yourself', 'yourselves', 'he', 'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itsel' 'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that 'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has 'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because' 'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'the 'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off 'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all' 'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've 've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "di 'hadn't", 'hasn', "hasn't", 'haven't", 'isn', "isn't", 'ma', 'mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn' 'won', "won', "won't", 'wouldn', "wouldn't"]
```

```
In [38]: from bs4 import BeautifulSoup
         # Combining all the above stundents
         from tqdm import tqdm
         preprocessed reviews = []
         # tqdm is for printing the status bar
         for sentance in tqdm(df['text'].values):
             sentance = re.sub(r"http\S+", "", sentance)
             sentance = BeautifulSoup(sentance, 'lxml').get text()
             sentance = decontracted(sentance)
             sentance = re.sub("\S*\d\S*", "", sentance).strip()
             sentance = re.sub('[^A-Za-z]+', ' ', sentance)
             # https://gist.github.com/sebleier/554280
             sentance = ' '.join(e.lower() for e in sentance.split() if e.lower() not in s
             preprocessed reviews.append(sentance.strip())
         100%| 75170/75170 [00:39<00:00, 1919.04it/s]
```

```
In [0]: | df['CleanedText']=preprocessed_reviews
```

[5] Model

```
In [0]: from keras.models import Model
        from keras.layers import Dense, Embedding, Input
        from keras.layers import LSTM, Bidirectional, GlobalMaxPool1D, Dropout
        from keras.preprocessing import text, sequence
        from keras.callbacks import EarlyStopping, ModelCheckpoint
```

```
In [0]:
        list sentences train = df["CleanedText"].values
        list_classes = ['Assembly Required', 'Color', 'Commercial / Residential',
                'ENERGY STAR Certified', 'Features', 'Finish',
                'Flooring Product Type', 'Hardware Included', 'Included',
                'Indoor/Outdoor', 'Package Quantity', 'Shape',
                'Tools Product Type', 'Voltage (volts)', 'Wattage (watts)']
        y = df[list classes].values
        list_sentences_test = test["text"].values
```

```
In [0]: #Standard keras preprocessing,
        #to turn each comment into a list of word indexes of equal length (with truncation
        max features = 20000
        tokenizer = text.Tokenizer(num_words=max_features)
        tokenizer.fit on texts(list(list sentences train))
        list tokenized train = tokenizer.texts to sequences(list sentences train)
        list tokenized test = tokenizer.texts to sequences(list sentences test)
        X train = sequence.pad sequences(list tokenized train, maxlen=maxlen)
        X test = sequence.pad sequences(list tokenized test, maxlen=maxlen)
```

```
#Simple bidirectional LSTM with two fully connected layers.
#adding some dropout to the LSTM since even 2 epochs is enough to overfit.
maxlen = 100
embed_size = 128
batch_size = 32
epochs = 2
inp = Input(shape=(maxlen, ))
x = Embedding(max_features, embed_size)(inp)
x = Bidirectional(LSTM(50, return_sequences=True))(x)
x = GlobalMaxPool1D()(x)
x = Dropout(0.1)(x)
x = Dense(50, activation="relu")(x)
x = Dropout(0.1)(x)
x = Dense(15, activation="sigmoid")(x)
model = Model(inputs=inp, outputs=x)
print(model.summary())
```

Output	Shape	Param #
(None,	100)	0
(None,	100, 128)	2560000
(None,	100, 100)	71600
(None,	100)	0
(None,	100)	0
(None,	50)	5050
(None,	50)	0
(None,	15)	765
	(None, (None, (None, (None, (None, (None,	Output Shape

Total params: 2,637,415 Trainable params: 2,637,415 Non-trainable params: 0

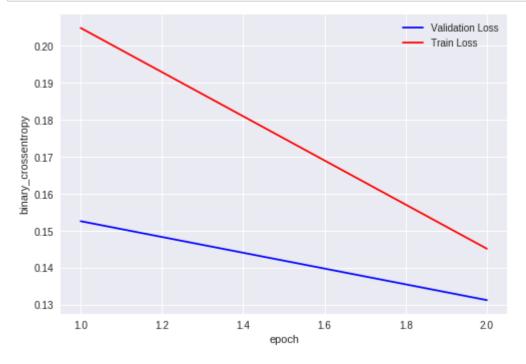
None

```
In [48]: # Compiling the model
        model.compile(loss='binary crossentropy',optimizer='adam',metrics=['accuracy'])
        # Fitting the data to the model
        history = model.fit(X_t, y, batch_size=batch_size, epochs=epochs, validation_spli
        Train on 52619 samples, validate on 22551 samples
        Epoch 1/2
        cc: 0.9309 - val loss: 0.1526 - val acc: 0.9385
        Epoch 2/2
        52619/52619 [============== ] - 445s 8ms/step - loss: 0.1452 - a
        cc: 0.9389 - val loss: 0.1313 - val acc: 0.9430
In [0]:
        import matplotlib.pyplot as plt
        import numpy as np
        import time
        # https://stackoverflow.com/a/14434334
        # this function is used to update the plots for each epoch and error
        def plt_dynamic(x, vy, ty, ax, colors=['b']):
            ax.plot(x, vy, 'b', label="Validation Loss")
            ax.plot(x, ty, 'r', label="Train Loss")
            plt.legend()
            plt.grid(True)
            fig.canvas.draw()
```

```
In [50]: # Validation plot
    fig,ax = plt.subplots(1,1)
    ax.set_xlabel('epoch'); ax.set_ylabel('binary_crossentropy')

# List of epoch numbers
    x = list(range(1,epochs+1))

vy = history.history['val_loss']
    ty = history.history['loss']
    plt_dynamic(x, vy, ty, ax)
```



```
In [52]: y_test = model.predict([X_test], batch_size=1024, verbose=1)
    sample_submission = pd.read_csv(f'sample_submission.csv')
    sample_submission[list_classes] = y_test
    sample_submission.to_csv('submission.csv', index=False)
```

10393/10393 [===========] - 13s 1ms/step

```
In [53]: b = pd.read_csv('submission.csv')
b.head()
```

Out[53]:

	id	Indoor/Outdoor	Commercial / Residential	ENERGY STAR Certified	Hardware Included	Package Quantity	Flooring Product Type	Color	Tool Produc Typ
0	114689	0.203735	0.001157	0.001561	0.000971	0.707016	0.000266	0.004056	0.00063
1	183172	0.002523	0.001735	0.001392	0.000128	0.001788	0.000316	0.000690	0.60553
2	217304	0.026385	0.000549	0.002213	0.011359	0.825187	0.000073	0.029446	0.01056
3	184115	0.331808	0.369900	0.000372	0.000119	0.001166	0.288054	0.015582	0.00035
4	103786	0.071457	0.065183	0.001432	0.029298	0.016845	0.006179	0.585410	0.00087
4									>

```
In [0]: from google.colab import files
  files.download('submission.csv')
In [0]:
```