

E6893 Big Data Analytics:

Reviews Exploration & Visualization

Team Members:

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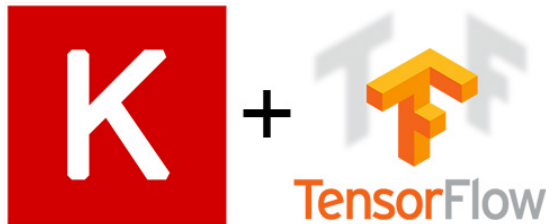
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- **Technologies Used**
- **Datasets Overview**
- **Challenges faced**
- **Data explore**
- **Review Stars Prediction**
- **Demo**
- **Conclusion**
- **Next Steps**

- **Language:** Python
- **Platform:** Apache Spark
- **Library:**
 - **ML :** NLTK, ScitKitLearn, Keras, TensorFlow
 - **Visualization:** Seaborn



- **Dataset Provided by Yelp**

The Challenge Dataset:

- **2.7M** reviews and **649K** tips by **687K** users for **86K** businesses
- **566K** business attributes, e.g., hours, parking availability, ambience.
- Social network of **687K** users for a total of **4.2M** social edges.
- Aggregated check-ins over time for each of the **86K** businesses
- **200,000** pictures from the included businesses

- **Included Five JSON File (Total 2.8G) :**

- **User Information**
- **Business Information**
- **Tips (text)**
- **Reviews**
- **Check-in Details**

review

```
{
  'type': 'review',
  'business_id': (encrypted business id),
  'user_id': (encrypted user id),
  'stars': (star rating, rounded to half-stars),
  'text': (review text),
  'date': (date, formatted like '2012-03-14'),
  'votes': {(vote type): (count)},
}
```

- DataSets are not very organized:
 - Some Fields are empty: eg. Price Range, Noise Level...

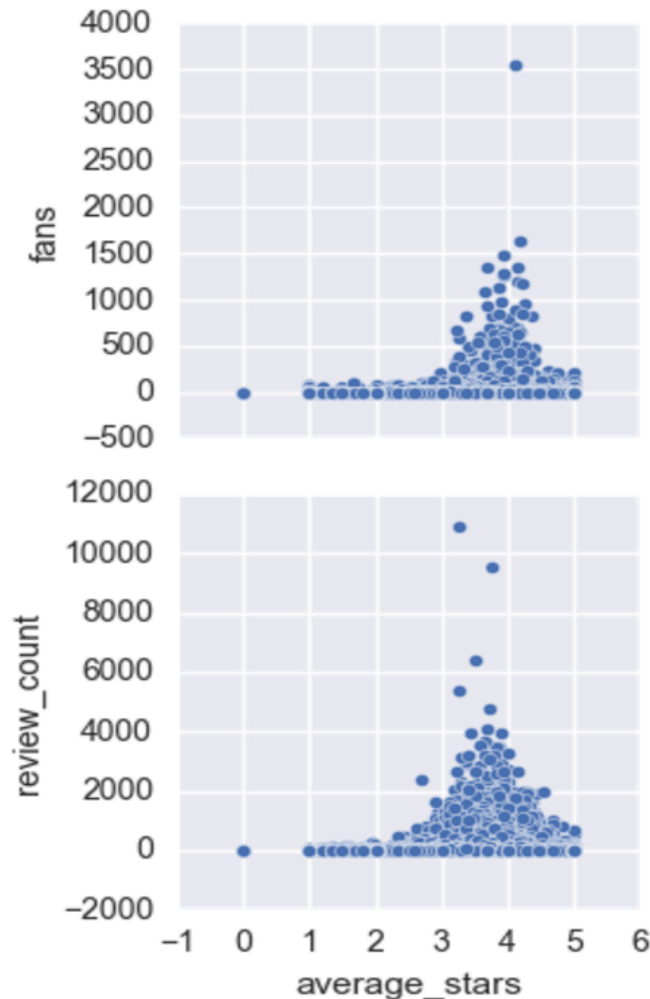
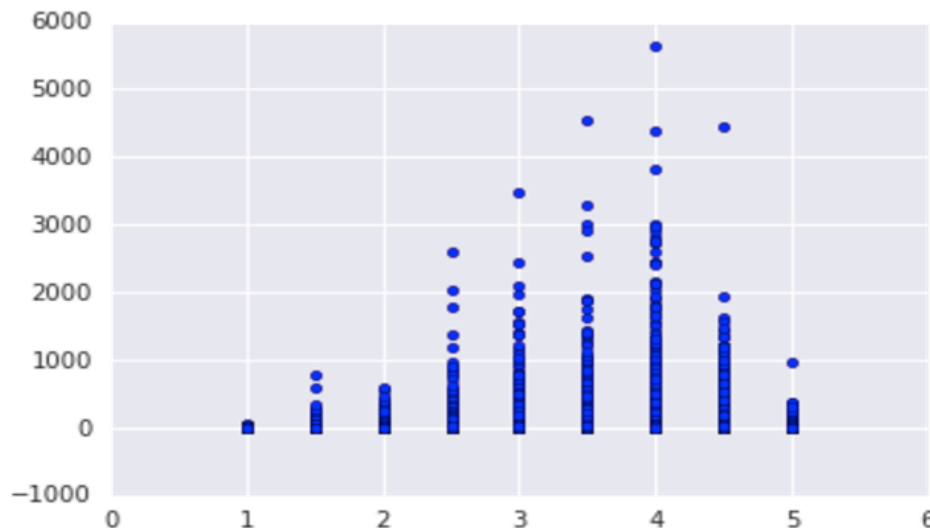


| Noise Level |
|-------------|
| loud |
| |
| |
| quiet |
| average |

- Some Fields are not easy to normalize: eg. WIFI: {no, free, paid, yes, ...}
- Datasets Categories are not well balanced (bootstrap resampling)

- **Natural Language Processing**
- **Sentiment Analysis**
- **Visual Analysis**
- **Feature Importance**

- Sentiments are more obvious in the 'tips' text as opposed to the 'reviews' text.
- Most users average ratings are between 3.5 - 4.5 stars.
- Most businesses average ratings are between 3.5 - 4.5 stars.



- Use word2vec as the word embedding layer
- Transform each review into a fixed length of words with each word represented by its word2vec vector
 - max number of words for each review : 50
 - max number of word features in the word2vec model : 5000
- The architecture of this model:
Embedding layer - Dropout - Convolution1D - MaxPooling1D - Full Connected layer - Dropout - Relu activation - Sigmoid (with binary cross entropy loss)
- Accuracy: training accuracy: 67.28%, validation accuracy: 67.21%

Train on 800000 samples, validate on 200000 samples

Epoch 1/2

800000/800000 [=====] - 798s - loss: 0.6258 - acc: 0.6728 - val_loss: 0.6215 - val_acc: 0.6721

Epoch 2/2

800000/800000 [=====] - 771s - loss: 0.6223 - acc: 0.6729 - val_loss: 0.6233 - val_acc: 0.6721

- Text preprocessing: remove stopwords, using bigrams model
- Utilize sentiment analysis result into stars prediction
- Bigram Multinomial Bayes Classifier

MODEL: Random Forest (100 Learners) with 20% Training Data

Precision: 0.7175643803

Recall: 0.72508315773

F1: 0.710133845734

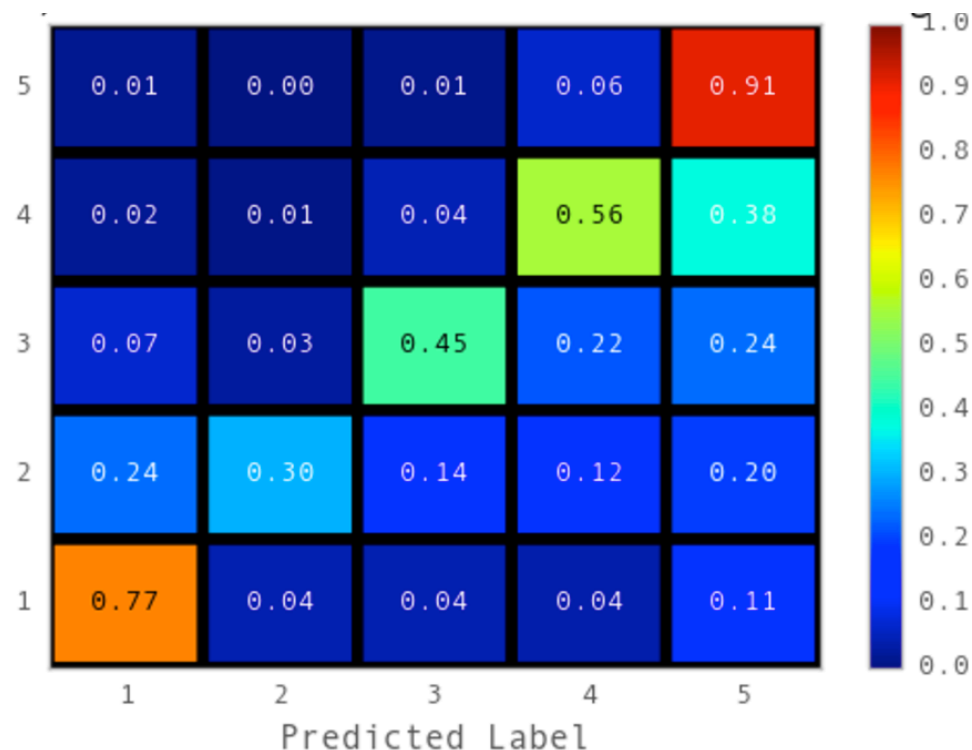
Accuracy: 0.72508315773

Classification Report:

| | precision | recall | f1-score | support |
|-------------|-----------|--------|----------|---------|
| 1 star | 0.63 | 0.77 | 0.69 | 3869 |
| 2 star | 0.63 | 0.30 | 0.41 | 2777 |
| 3 star | 0.62 | 0.45 | 0.52 | 5712 |
| 4 star | 0.72 | 0.56 | 0.63 | 16248 |
| 5 star | 0.76 | 0.91 | 0.83 | 28815 |
| avg / total | 0.72 | 0.73 | 0.71 | 57421 |

Precision variance: 0.014963

Recall variance: 0.238742



- **Use visualization technique to explore the data, get the overview of the structure of the data, generate wordcloud, identify important features.**
- **Use machine learning algorithms to perform sentiment analysis and predict the reviews stars**
- **Build web pages to let user explore restaurant reviews interactively.**

- **There are various interesting questions can be explored on the yelp datasets !!!**
- **Graph Mining:** Figure out who the trend setters are? How much influence does people's social circle have on their business choices and their ratings?
- **Seasonal Trends:** Are there more reviews for sports bars on major game days and if so, could you predict that?
- **Location Mining and Urban Planning:** How much of a business' success is really just location, location, location? Do you see reviewers' behavior change when they travel?

