Yelp Data Prediction

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Monday Group 1

1 Introduction and Data Cleaning

Introduction

- Small set of informative features
- Accurate predictive model
- Based on about 1.5 million Yelp reviews

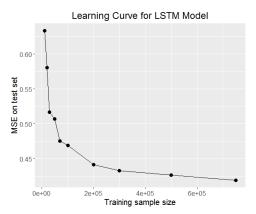
Data Cleaning

- Modify Abbreviation and Special Symbol
- Remove Non-English
- Negative Sentences
- Remove Punctuation

2 Model: LSTM

Model: Neural Network with 3 layers

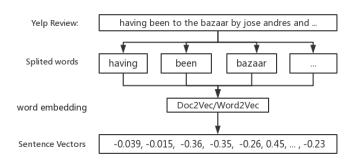
- layer1: LSTM layer with 50 output units
- layer2: Dense layer with 3 output units
- layer3: Dense layer with 1 output unit



2.1 Pretrained Sentence Vectores

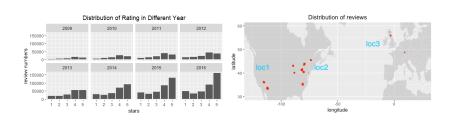
Model Features

- Pre-trained Sentence Vectors: Capture word counts and order
- Additional Variables: Capture sentiment, review date and location



2.2 Additional Variables

- year: scaled year variable.
- loc1: 1 if the restaurant is in the western United States, otherwise 0.
- loc2: 1 if the restaurant is in the eastern United States, otherwise 0.



2.2 Additional Variables

- **Score1** \sim 5: Score1[word] = $\frac{P(\text{this word is included in reviews with 1-star})}{P(\text{this word is included in reviews with other stars})}$
- ${\bf S1}\sim {\bf S5}$: ${\bf S1}[{\sf review}]=\#$ of words with high Score1 in the review.

| Word | Variable | 1-star | 2-star | 3-star | 4-star | 5-star |
|----------------|-------------|--------|--------|--------|--------|--------|
| refund | frequence | 115 | 15 | 7 | 4 | 2 |
| | probability | 0.011 | 0.002 | 0 | 0 | 0 |
| | Score | 34.200 | 1.080 | 0.300 | 0.072 | 0.025 |
| notdisappoints | frequence | 0 | 2 | 5 | 43 | 110 |
| | probability | 0 | 0 | 0 | 0.002 | 0.003 |
| | Score | 0 | 0.116 | 0.188 | 0.917 | 3.870 |
| and | frequence | 9196 | 8691 | 12851 | 25604 | 32071 |
| | probability | 0.859 | 0.886 | 0.877 | 0.895 | 0.886 |
| | Score | 0.968 | 1.000 | 0.991 | 1.020 | 1.000 |

2.2 Additional Variables

Positive

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Negative

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3 Compare MSE with other method

MSE

| LM | NB | NN | LSTM | GLM | SVM |
|-------|--------------------------------------|--|--|---|---|
| 0.673 | 0.974 | 0.494 | 0.493 | 0.698 | NA |
| 0.720 | 1.112 | 0.524 | 0.526 | 0.756 | 0.585 |
| 0.836 | 1.459 | 0.614 | 0.612 | 0.894 | NA |
| NA | 1.126 | 1.210 | NA | 0.864 | 0.790 |
| 0.889 | 1.114 | 0.804 | NA | 0.836 | 0.770 |
| | 0.673 0.720 0.836 NA | 0.673 0.974 0.720 1.112 0.836 1.459 NA 1.126 | 0.673 0.974 0.494 0.720 1.112 0.524 0.836 1.459 0.614 NA 1.126 1.210 | 0.673 0.974 0.494 0.493 0.720 1.112 0.524 0.526 0.836 1.459 0.614 0.612 NA 1.126 1.210 NA | 0.673 0.974 0.494 0.493 0.698 0.720 1.112 0.524 0.526 0.756 0.836 1.459 0.614 0.612 0.894 NA 1.126 1.210 NA 0.864 |

tested on 100000 data

4 Interpretable Model

$$\hat{y} = 3.65 + 0.04 * scale(year) + 0.04 * loc1 + 0.06 * loc2 - 0.11 * S1 - 0.17 * S2 - 0.03 * S3 + 0.03 * S4 + 0.14 * S5$$

5 Strengths and Weaknesses

Strengths

MSE 0.493 for best model feature combination prediction Inclusion of additional informative variables contributes to the reduction of MSE by 0.033

Weaknesses

Grid search over various model parameters

Thank You!