# Image Help From Yelp

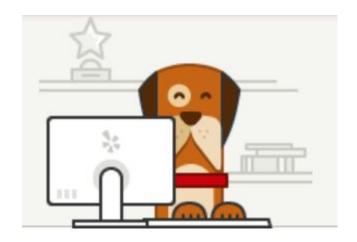
Insights into Yelp's Open Dataset
Feb 2020

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Github.com/yuchild/image\_help\_from\_yelp

# **Case Study Questions**

- 1. Can written reviews *predict* ratings?
- Can the photos taken be classified?
- 3. Can **photos** taken help us **rate** an establishment?







#### Data Source: Yelp Open Dataset











192,609 businesses

200,000 pictures

10 metropolitan areas

1,223,094 tips by 1,637,138 users

Over 1.2 million business attributes like hours, parking, availability, and ambience Aggregated check-ins over time for each of the 192,609 businesses

Source: yelp.com/dataset

# Summary Yelp Open Dataset Used:





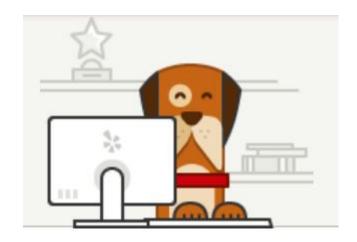




File Name	Number of Entries	Attributes
business.json	192609	names, stars, reviews_count, city, state, attributes, categories
checkin.json	161950	business_id, dates
photo.json	200000	caption, label
review.json	5376719	review_id, user_id, business_id, stars, useful, funny, cool, text, date
tip.json	1223094	text, date, compliment_count
user.json	1637138	review_count, useful, funny, cool, fans, avg_stars, compliment_hot

#### **Quick Overview**

- 1. Written reviews predicts ratings with 85% accuracy
- 2. Classification of **images** was problematic with **SVM** with **45%** accuracy
- 3. Photos classify business ratings with **62%** accuracy using CNN







#### Reviews

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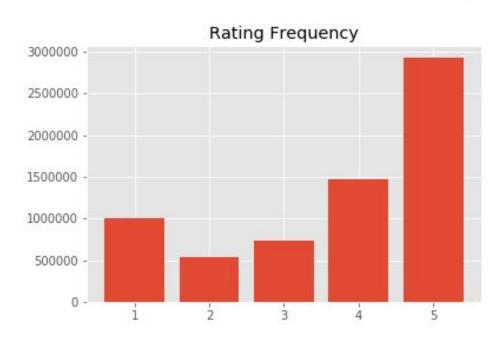


3 (6.	AL
.0 Total bill for this horrible service? Over \$8G	j
.0 Today was my second out of three sessions I ha	L
O This place has gone down hill. Clearly they h	۱

text

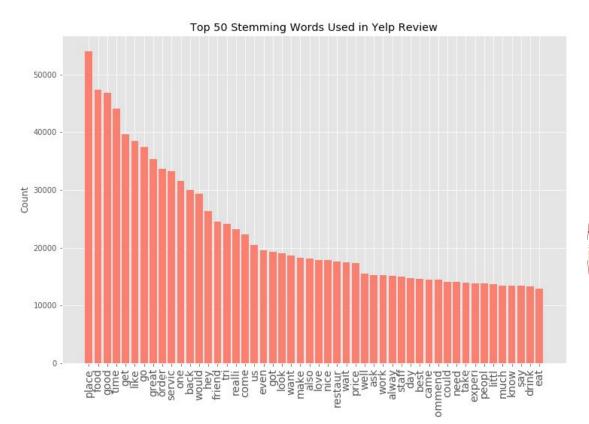
text	stars
sert had a big name in Hong Kong and	3.0
a giant Best Buy with 66 registers. I do	3.0
hinese food and I love mexican food. W	3.0

stars	te te	
5.0	I *adore* Travis at the Hard Rock's new Kelly	
5.0	I have to say that this office really has it t	
5.0	Went in for a lunch. Steak sandwich was delici	





# **NLP Top Stemming Words**









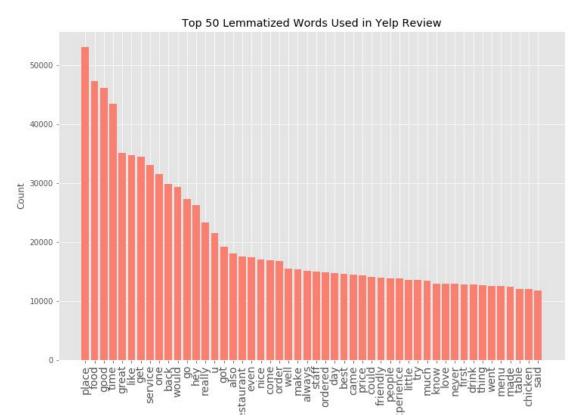








# **NLP Top Lemmatized Words**











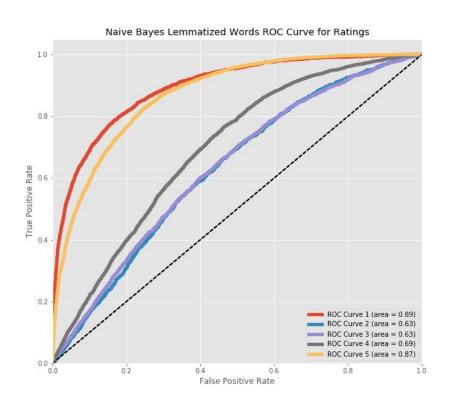


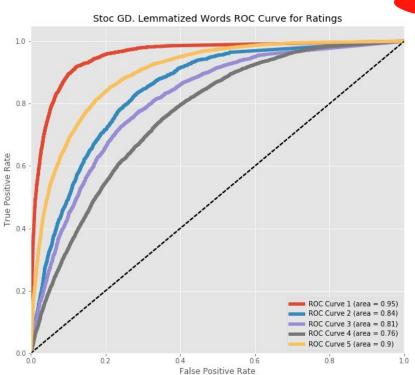




#### Clash of the Models: **NB** vs **SVM** (SGD)



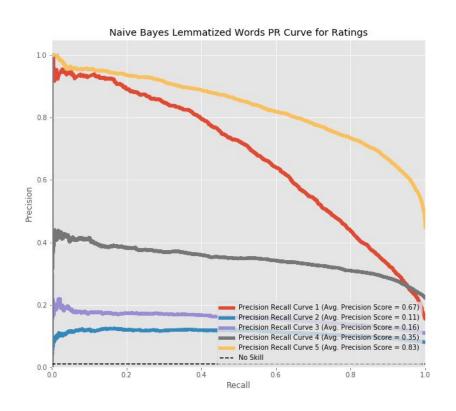


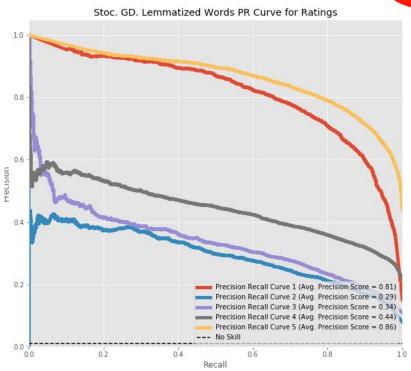




### Clash of the Models: **NB** vs **SVM** (SGD)









#### Text Ratings Takeaways:

- 1. Written reviews prediction with 85% accuracy
- 2. Model is **biased** towards the two ends of the scale: **1** and **5** stars
- 3. Ratings **2**, **3**, and **4** stars poorly classified even with *balanced* training set
- 4. Only extreme words will trigger a poor rating from the model

s te	stars	S
0 Wish I could give this place 0 stars. We have	1.0	
0 I didn't listen to the low reviews, I wish I d	1.0	
0 Below average food. The service can be spotty,	1.0	







Can a picture tell you anything about the establishment?



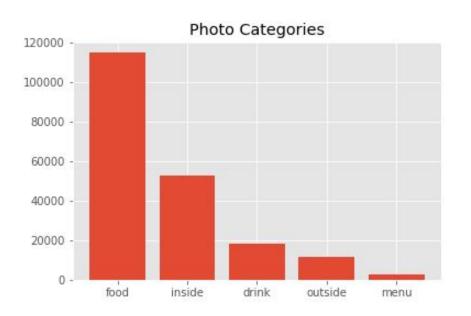


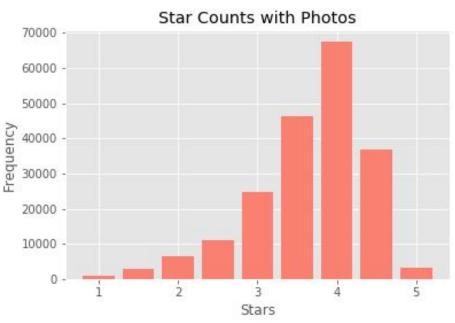


#### **Photo Classification**





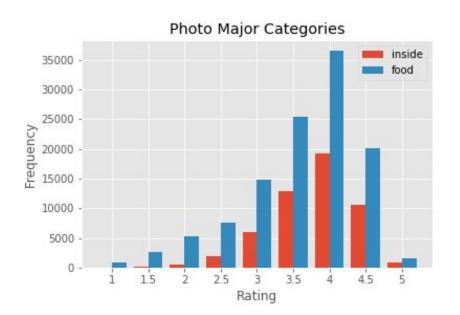


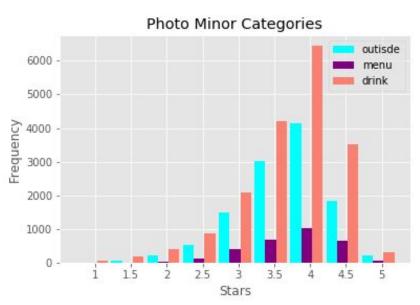


#### Photos of What?









### Photo Classifier **SVM** (SGD):





5 fold Cross Validated at 31% accuracy

```
['menu', 'inside', 'food', ..., 'food', 'drink', 'menu']
['food', 'food', 'food', ..., 'drink', 'menu', 'menu'],
```

pred\_img(file\_11)

#### Photo Classifier **SVM** (SGD):

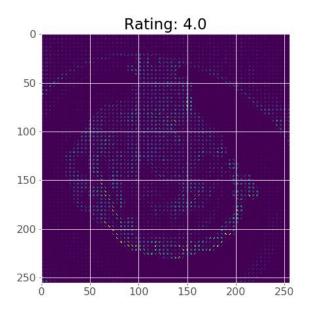
5 fold Cross Validated at 31% accuracy

HOG'in the image increased accuracy to 42%





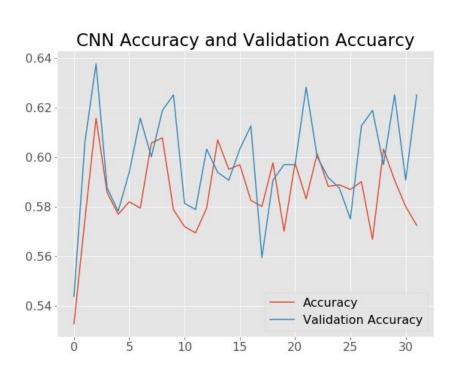




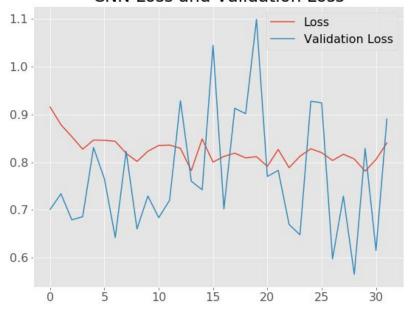
#### **CNN** To The Rescue







#### **CNN Loss and Validation Loss**

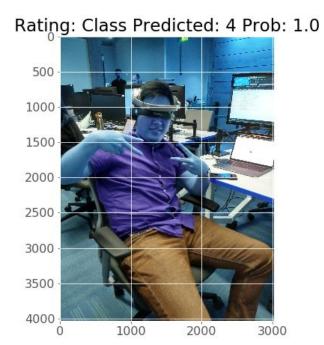


# CNN for Fun, Maybe Not So Much









#### Conclusion





**SVM** turned to be better at **text** to rating classification than photos

CNN is better at classifying photos than SVM by 20%

Future work includes CNN tuning with more HOG or other image preprocessing

for a better model

Models are **NOT** operational and will need future retooling

