

Target Audience

If you have done any model training in the past, this workshop is likely too basic for you.



Before The Workshop

- 1. Download the dataset
- download from Yelp Open Dataset and choose the json format
- unzip to your desired folder: tar -xzvf yelp_dataset.tar
- 2. Install python in your local computer and all the required libraries, via Anaconda
- download from Anaconda site and choose the version for python 3.7
- note all libraries we need in the workshop are pre-packaged in Anaconda so you don't need to pip install bluh

How You Can Spend the Next Hour

You can directly run the notebook on your computer if you have the dataset downloaded and environment set up. Don't worry if you don't.

We will make this workshop as interactive as possible by walking through my slides (40+ pages) full of screenshots of the notebook, and explain what each step is about. Feel free to raise your hand to ask questions any time during the workshop.

Github ipynb <u>renderer</u>.



Yelp Open Dataset yelp.com/dataset



6,685,900 reviews 192,609 businesses

How to run the notebook in your browser

• clone the repo

git clone https://github.com/xun-tang/ds_workshop.git

• run the anaconda-navigator app Anaconda-Navigator



- navigate to the folder of checked-out repo
- double click to launch the notebook file ending with .ipynb

Why not run in <u>y/jupyterhub</u> directly?



Will Your Next **yelp** Review

Be a **Mathe Review**?



Given a user and a business, predict whether the user will write a Carlow Carlow for the business.







Unsupervised learning







"Learn this function"

Task

Classification

Regression Ranking

Labels (function output)

Binary (boolean) or categorical

Continuous (float) value 📈

Ordered items

To Solve a Data Science Problem

- Step 1: Load the Data
- Step 2: Explore and Visualize the Data
- Step 3: Generate the Features
- Step 4: Train a Model
- Step 5: Evaluate the Model
- Step 6 & Beyond: Iterate Through the Process

Step 1 Load the Data



Store Data in Pandas DataFrames

	In [1]:	%%time import pandas as pd
		CHUNK_SIZE = 10000 PATH = '~/Desktop/yelp_dataset/'
		CPU times: user 407 ms, sys: 203 ms, total: 610 ms Wall time: 1.77 s
	In [2]:	<pre>%%time review_df = pd.concat(pd.read_json(PATH + 'review.json', lines=True, chunksize=CHUNK_SIZE)) review_df = review_df.set_index('review_id')</pre>
		CPU times: user 1min 26s, sys: 19.1 s, total: 1min 45s Wall time: 1min 46s
★国	In [3]:	<pre>%%time user_df = pd.concat(pd.read_json(PATH + 'user.json', lines=True, chunksize=CHUNK_SIZE)) user_df = user_df.set_index('user_id')</pre>
		CPU times: user 40.9 s, sys: 10.5 s, total: 51.4 s Wall time: 50.9 s
	In [4]:	<pre>%%time biz_df = pd.concat(pd.read_json(PATH + 'business.json', lines=True, chunksize=CHUNK_SIZE)) biz_df = biz_df.set_index('business_id')</pre>
		CPU times: user 5.84 s, sys: 1.05 s, total: 6.89 s Wall time: 6.04 s

What's in a Review DataFrame?

review_df.head(): Print top rows in the data frame.

review_df.describe(): Generate various summary statistics, mean, max, count, etc.

funny	user_id	text	business_id	stars	useful	type	cool	datetime	year
0	KpkOkG6Rlf4Ra25Lhhxf1A	If you enjoy service by someone who is as comp	2aFiy99vNLklCx3T_tGS9A	5	0	review	0	2011-10- 10	2011
0	bQ7fQq1otn9hKX- gXRsrgA	After being on the phone with Verizon Wireless	2aFiy99vNLklCx3T_tGS9A	5	1	review	0	2010-12- 29	2010
0	r1NUhdNmL6yU9Bn- Yx6FTw	Great service! Corey is very service oriented	2aFiy99vNLkICx3T_tGS9A	5	0	review	0	2011-04- 29	2011
0	aW3ix1KNZAvoM8q- WghA3Q	Highly recommended. Went in yesterday looking	2LfluF3_sX6uwe-IR-P0jQ	5	0	review	1	2014-07- 14	2014
0	YOo- Cip8HqvKp_p9nEGphw	I walked in here looking for a specific	2LfluF3_sX6uwe-IR-P0jQ	4	0	review	0	2014-01- 15	2014
	0 0 0 0 0 0 0 0 0	0 KpkOkG6Rlf4Ra25Lhhxf1A 0 bQ7fQq1otn9hKX- gXRsrgA 0 r1NUhdNmL6yU9Bn- Yx6FTw 0 aW3ix1KNZAvoM8q- WghA3Q 0 YOo- Cin8HquKp, pqnEGphw,	0 KpkOkG6RIf4Ra25Lhhxf1A If you enjoy service by someone who is as comp 0 bQ7fQq1otn9hKX- gXRsrgA If you enjoy service by someone who is as comp 0 bQ7fQq1otn9hKX- gXRsrgA After being on the phone with Verizon Wireless 0 r1NUhdNmL6yU9Bn- Yx6FTw Great service! Corey is very service oriented 0 aW3ix1KNZAvoM8q- WghA3Q Highly recommended. Went in yesterday looking 0 YOo- Cio8HouKo pgnEGobw I walked in here looking	0 KpkOkG6RIf4Ra25Lhhxf1A If you enjoy service by someone who is as comp 2aFiy99vNLklCx3T_tGS9A 0 bQ7fQq1otn9hKX- gXRsrgA After being on the phone with Verizon 2aFiy99vNLklCx3T_tGS9A 0 r1NUhdNmL6yU9Bn- Yx6FTw Great servicel Corey is very service oriented 2aFiy99vNLklCx3T_tGS9A 0 r1NUhdNmL6yU9Bn- Yx6FTw Great servicel Corey is very service oriented 2aFiy99vNLklCx3T_tGS9A 0 aW3ix1KNZAvoM8q- WghA3Q Highly recommended. Went in yesterday looking 2aFiy99vNLklCx3T_tGS9A 0 YOo- Cin8HcwKp_pgnEGphw I walked in here looking for a service 2LfluF3_sX6uwe-IR-P0jQ	0 KpkOkG6RIf4Ra25Lhhxf1A If you enjoy service by someone who is as comp 2aFiy99vNLklCx3T_tGS9A 5 0 bQ7fQq1otn9hKX- gXRsrgA After being on the phone with Verizon Wireless 2aFiy99vNLklCx3T_tGS9A 5 0 r1NUhdNmL6yU9Bn- Yx6FTw Great service! Corey is very service oriented 2aFiy99vNLklCx3T_tGS9A 5 0 aW3ix1KNZAvoM8q- WghA3Q Highly recommended. Went in yesterday looking 2LfluF3_sX6uwe-IR-P0jQ 5 0 YOo- Cio8HouKo pgeEGobw I walked in here looking a specific 2LfluF3_sX6uwe-IR-P0jQ 4	0 KpkOkG6RIf4Ra25Lhhxf1A If you enjoy service by someone who is as comp 2aFiy99vNLklCx3T_tGS9A 5 0 0 bQ7fQq1otn9hKX- gXRsrgA After being on the phone with Verizon Wireless 2aFiy99vNLklCx3T_tGS9A 5 1 0 r1NUhdNmL6yU9Bn- Yx6FTw Great servicel Corey is very service oriented 2aFiy99vNLklCx3T_tGS9A 5 1 0 aW3ix1KNZAvoM8q- WghA3Q Highly recommended. Went in yesterday looking 2LfluF3_sX6uwe-IR-P0jQ 5 0 0 YOo- CioBiHouKo pgnEGohw I walked in here looking a spacific 2LfluF3_sX6uwe-IR-P0jQ 4 0	0 KpkOkG6RIf4Ra25Lhhxf1A If you enjoy service by someone who is as comp 2aFiy99vNLklCx3T_tGS9A 5 0 review 0 bQ7fQq1otn9hKX- gXRsrgA After being on the phone with Verizon Wireless 2aFiy99vNLklCx3T_tGS9A 5 1 review 0 r1NUhdNmL6yU9Bn- Yx6FTw Great servicel Corey is very service oriented 2aFiy99vNLklCx3T_tGS9A 5 0 review 0 aW3ix1KNZAvoM8q- WghA3Q Highly recommended. Went in yesterday looking 2LfluF3_sX6uwe-IR-P0jQ 5 0 review 0 YOo- CiosHowKo ponEGobw I walked in here looking 2LfluF3_sX6uwe-IR-P0jQ 4 0 review	0 KpkOkG6RIf4Ra25Lhhxf1A If you enjoy service by someone who is as comp 2aFiy99vNLklCx3T_tGS9A 5 0 review 0 0 bQ7fQq1otn9hKX- gXRsrgA After being on the phone with Verizon Wireless 2aFiy99vNLklCx3T_tGS9A 5 1 review 0 0 r1NUhdNmL6yU9Bn- Yx6FTw Great service! Corey is very service oriented 2aFiy99vNLklCx3T_tGS9A 5 1 review 0 0 aW3ix1KNZAvoM8q- WghA3Q Highly recommended. Went in yesterday looking 2LfluF3_sX6uwe-IR-P0jQ 5 0 review 1 0 YOo- CioBidrukko pgeEGobw I walked in here looking can specific 2LfluF3_sX6uwe-IR-P0jQ 4 0 review 0	0 KpkOkG6RIf4Ra25Lhhxf1A If you enjoy service by someone who is as comp 2aFiy99vNLklCx3T_tGS9A 5 0 review 0 2011-10-10 0 bQ7fQq1otn9hKX- gRsrgA After being on the phone with Verizon Wireless 2aFiy99vNLklCx3T_tGS9A 5 1 review 0 2011-12-29 0 r1NUhdNmL6yU9Bn- Yx6FTw Great servicel Corey is very service oriented 2aFiy99vNLklCx3T_tGS9A 5 1 review 0 2011-04-29 0 aW3ix1KNZAvoM8q- WghA3Q Highly recommended. Went in yesterday looking 2LfluF3_sX6uwe-IR-P0jQ 5 0 review 1 2014-07-14 0 YOo- Cip8HorkKo pgnEGobw I walked in here looking 2LfluF3_sX6uwe-IR-P0jQ 4 0 review 0 2014-01-16-16

Step 2 Explore and Visualize the Data



Review Star Rating Distribution Published on Yelp's Factsheet



Source: <u>https://www.yelp.com/factsheet</u>



Plot Review Star Rating Distribution from Open Dataset import seaborn as sns

%matplotlib inline

ax = sns.countplot(x='stars', data=review df)



Plot Star Ratings by Year

ax = sns.countplot(x='year', data=review_df, hue='stars')



Step 3 Generate the Features



For Example..

Convert date string to date delta

e.g. business_age

Convert strings to categorical features

e.g. noise level: {`quiet', `loud', `very loud'}.

Drop unused features

e.g. business_name

compute days in between date and max value in date
def calculate_date_delta(df, column):
 to_column = column + '_delta'
 datetime = pd.to_datetime(df[column])
 time_delta = datetime.max() - datetime
 df[to_column] = time_delta.apply(lambda x: x.days)
 df.drop(column, axis=1, inplace=True)

compute length of string

def to_length(df, column): to_column = column + '_len' df[to_column] = df[column].apply(lambda x: len(x)) df.drop(column, axis=1, inplace=True)

def drop_columns(df, columns):
 for column in columns:
 df.drop(column, axis=1, inplace=True)

def to_boolean(df, columns):
 for column in columns:
 to_column = column + '_bool'
 df[to_column] = df[column].apply(lambda x: bool(x)))
 df.drop(column, axis=1, inplace=True)

FILL_WITH = 0.0

def to_category(df, columns):
 for column in columns:
 df[column] = df[column].astype('category')
 # add FILL_WITH category for fillna()
 if (FILL_WITH not in df[column].cat.categories):
 df[column] = df[column].cat.add_categories([FILL_WITH])
 print(f'categories for {key} include {df[key].cat.categories}')

def category_rename_to_int(df, columns):
 for column in columns:
 df[column].cat.remove_unused_categories()
 size = len(df[column].cat.categories)
 print(f'column {column} has {size} columns, including {df[column].cat.categories}')
 df[column] = df[column].cat.remme_categories(range(1, size+1))
 print(f'=> {df[column].cat.categories}')

Join DataFrames to Populate the Features



The `user_df` DataFrame is already indexed by the join key (`user_id`). Make sure it's on the right side of join.
review_join_user = review_df.join(user_df, on='user_id')

review_join_user_join_biz = review_join_user.join(biz_df, on='business_id')



Step 4 Train a Model



Arrange Data into a Feature Matrix and a Target Array

Feature matrix (X)

All features generated from biz, user, review dataframes

Target array (y)

What we predict: Whether the review is Five-star or not

Target y is whether a review is five-star (True / False)
y = review_join_user_join_biz.review_stars.apply(lambda x: x == 5)

Exclude the `stars` columns from the feature matrix, since it is the target
X = review_join_user_join_biz
review_join_user_join_biz.drop('review_stars', axis=1, inplace=True)

Split Training and Testing Set

Training set: used for an machine learning algorithm to train from

Testing set: used to to estimate / evaluate how well the model has been trained

Split them s.t. we don't evaluate on the same dataset we train from

from sklearn.cross_validation import train_test_split

Split the data into a training set and a test set
X_train, X_test, y_train, y_test = train_test_split(X, y)

training data shape (3552672, 109)
test data shape (1184225, 109)
converted label data shape (3552672,)



What Model to Use?



Choose the model to use

Logistic Regression (LR)

Estimates the prob. of a **binary** response based on the features

Here we estimate the prob. of a review being five-star

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from sklearn import linear_model

Build model using default parameter values
lrc = linear_model.LogisticRegression()

Normalize the Features

Standardize features by removing the mean and scaling to unit variance Logistic Regression requires all features normalized

from sklearn import preprocessing

scaler = preprocessing.StandardScaler().fit(X_train)

X_train_scaled = scaler.transform(X_train)
X test scaled = scaler.transform(X test)



Modeling

Key methods:

fit: Fit the model according to the given training data predict: Predict class labels for samples in data score: Returns the mean accuracy on the given test data and labels



from sklearn import linear_model

Build model using default parameter values
lrc = linear model.LogisticRegression(solver='lbfgs')

```
%%time
lrc_fit = lrc.fit(X_train_scaled, y_train)
y_pred = lrc_fit.predict(X_test_scaled)
```

```
CPU times: user 2min 20s, sys: 4.82 s, total: 2min 25s
Wall time: 27.4 s
```



Step 5 Evaluate the Model



Given users' past reviews on Yelp

Does It Work?

When the user writes a review for a business she hasn't reviewed before

Will it be a 🔂 🔂 🔂 🔂 review? (True / False)



Given users' past reviews on Yelp

user1 = user_df[user_df.index == 'kEtR1ZVL3Xr-tEX7lg16dQ']
#print user1.review_count
print user1.average stars

user_id
kEtR1ZVL3Xr-tEX7lg16dQ 4.96
Name: average stars, dtype: float64

user2 = user_df[user_df.index == 'Hj20fg3vyzKnJwnLn_rMqw']
#print user2.review_count
print user2.average_stars

user_id
Hj20fg3vyzKnJwnLn_rMqw 4.55
Name: average stars, dtype: float64

user3 = user_df[user_df.index == 'om5ZiponkpRqUNa3pVPiRg']
#print user2.review_count
print user3.average stars

user_id om5ZiponkpRqUNa3pVPiRg 3.94 Name: average_stars, dtype: float64



When the user writes a review for a business she hasn't reviewed before



Will it be a 🔀 🔂 🔂 🔂 review?

predict_given_user_biz(user=user1, biz=biz1, review_df=review_df)
predict_given_user_biz(user=user2, biz=biz1, review_df=review_df)
predict_given_user_biz(user=user3, biz=biz1, review_df=review_df)

True, with probability [False, True] == [0.08 0.92] True, with probability [False, True] == [0.22 0.78] False, with probability [False, True] == [0.63 0.37]

Make predictions for user[1,2,3]'s review on biz2

predict_given_user_biz(user=user1, biz=biz2, review_df=review_df)
predict_given_user_biz(user=user2, biz=biz2, review_df=review_df)
predict_given_user_biz(user=user3, biz=biz2, review_df=review_df)

```
True, with probability [False, True] == [0.46 0.54]
False, with probability [False, True] == [0.75 0.25]
False, with probability [False, True] == [0.95 0.05]
```

Confusion Matrix



Normalized Confusion Matrix



Cross Validation

- Holding out a portion of the training data for model validation, and do this for K_FOLDS
 - Ensure that the model does not overfit the training data
 - Select optimal model parameters
- from sklearn.model_selection import cross_validate
 import numpy as np

```
# Function used to calculate and print cross-validation scores
def training_score(est, X, y, cv):
    scores = cross_validate(est, X, y, cv=cv, scoring=['accuracy', 'roc_auc'])
```

Accuracy

Percentage of labels correctly predicted. The higher the better.

training score(est=lrc, X=X_train_scaled, y=y_train, cv=K_FOLD) 5-fold Train Cross Validation Accuracy: 0.749 +/- 0.0 | ROC A



ROC AUC

ROC curve, is a graphical plot that illustrates the diagnostic ability of a binary classifier system as its discrimination threshold is varied. ROC AUC ("Area Under Curve"). The higher the better.



Step 6 & Beyond Iterate Through the Process

