

# NewYorker Data Science Exercise

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Github repo: [https://github.com/ck-unifr/yelp\\_dataset\\_challenge](https://github.com/ck-unifr/yelp_dataset_challenge)

# Case 1: predict star ratings with checkin data

- The goal is to predict the star ratings based on the business and checkin data.
- The problem is formulated as a multi-label classification problem.
- Steps [\*]
  - Load data into pandas dataframe
    - *business.json* contains the business data
    - *checkin.json* contains the check in data
  - Data preprocessing
    - Merge the business data and checkin data
    - Fill missing value
    - Transfer string features into integer
    - Normalize numeric features
    - Extract text feature with tf-idf
    - Data is splitted into train set and test set
  - Model training
    - Two ensemble learning models, i.e., random forest, gradient boost machine, are used
    - Random search is used for hyperparameter tuning

[\*] The details can be found in the jupyter notebook

[https://github.com/ck-unifr/yelp\\_dataset\\_challenge/blob/master/notebooks/predict\\_star\\_ratings\\_with\\_checkin.ipynb](https://github.com/ck-unifr/yelp_dataset_challenge/blob/master/notebooks/predict_star_ratings_with_checkin.ipynb)

# Case 1: predict star ratings with checkin data

- Conclusion [\*]
  - Gradient boost machine (xgboost) with hyperparameter tuning achieves slightly better performance than random forest.
  - tf-idf 'categories' does not improve the performance.
  - The difficulty is the imbalanced data, i.e., we have only one thousand samples with one star in our train set and we have much more samples with 4 stars.
- Future work
  - Error analysis.
  - One-hot encoding 'categories'.
  - Word Embedding 'categories'
  - Using sampling algorithms to overcome the imbalanced data problem.
    - For example, python imbalanced-learn (<https://imbalanced-learn.org/en/stable/index.html>)

[\*] The details can be found in the jupyter notebook

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# Case 2: predict star ratings with photo

- The goal is to predict the star ratings based on the photos
- The problem is formulated as a multi-label classification problem
- Steps [\*]
  - Load data into pandas dataframe [\*\*]
    - *business.json* contains the business data
    - *photo.json* contains the photo data
  - Data preprocessing
    - Merge the business and photo data
    - Prepare the image data, i.e., transfer image into array in python
  - Model training
    - Convolutional neural network is used for the classification task

[\*] The details can be found in the jupyter notebook

[https://github.com/ck-unifr/yelp\\_dataset\\_challenge/blob/master/notebooks/predict\\_star\\_ratings\\_with\\_photos.ipynb](https://github.com/ck-unifr/yelp_dataset_challenge/blob/master/notebooks/predict_star_ratings_with_photos.ipynb)

[\*\*] Due the the limited computational resource I have, only a part of the photos is used.

# Case 2: predict star ratings with photo

- Conclusion [\*]
  - The performance is better than using checkin data, however only a small part of data is used in this case. So the comparison is not fair.
  - The difficulty is the imbalanced data as mentioned in case 1.
  - Error analysis shows that even for human it is hard to predict stars by images
- Future work
  - Combine text data, 'caption'
  - Transfer learning with VGG neural network.
  - Try other CNNs, e.g., GoogleNet, ResNet, OctConv.
  - Using sampling algorithms to overcome the imbalanced data problem.
    - GANs

[\*] The details can be found in the jupyter notebook

[https://github.com/ck-unifr/yelp\\_dataset\\_challenge/blob/master/notebooks/predict\\_star\\_ratings\\_with\\_photos.ipynb](https://github.com/ck-unifr/yelp_dataset_challenge/blob/master/notebooks/predict_star_ratings_with_photos.ipynb)

# Docker

- Type the command in terminal
  - `docker run -it --rm --name ds-jupyter -p 8888:8888 -v [work directory]:/home/jovyan/work jupyter/datascience-notebook`
- Open a web browser and enter the address which is indicated in the terminal
  - For example
    - `http://127.0.0.1:8888/?token=5b02107f343bc4bab3d87e81a9821c70e20be879a9620983`