Lab. – Regression:

House Sale Price Prediction Challenge

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房價預測 (回歸模型)

任務

- 用train.csv跟valid.csv訓練模型(一行是一筆房屋交易資料的紀錄,包括id, price與21種房屋參數)
- 將test.csv中的每一筆房屋參數,輸入訓練好的模型,預測其房價
- 將預測結果上傳到Kaggle(從 "Submit Predictions"連結)
- 看系統幫你算出來的Mean Abslute Error(MAE,就是跟實際房價差多少,取絕對值)分數夠不夠好?
- 嘗試改進預測模型

討論

- 資料分析
- 做法
- 程式寫法
- 結果分析
- 檢討與改進

作業進行方式

- 使用Github與Kaggle的教室平台
 - 把結果放到Kaggle(submission)上排名
 - 把程式與報告放到Github(至少要有readme.md(報告), train.sh,test.sh,requirement.txt與報告)
- 以邀請碼(另外通知)連結申請GitHub帳號,並以"學號+姓名",當你的帳號名稱(名稱 需與Kaggle的帳號相同
 - 建帳號時要跟課程學生名單上你的學號連結
- 以邀請碼(另外通知)連結申請Kaggle帳號,並以"學號+姓名",當你的帳號名稱帳號 (名稱需與Github的帳號相同):
 - 與你的github帳號關連

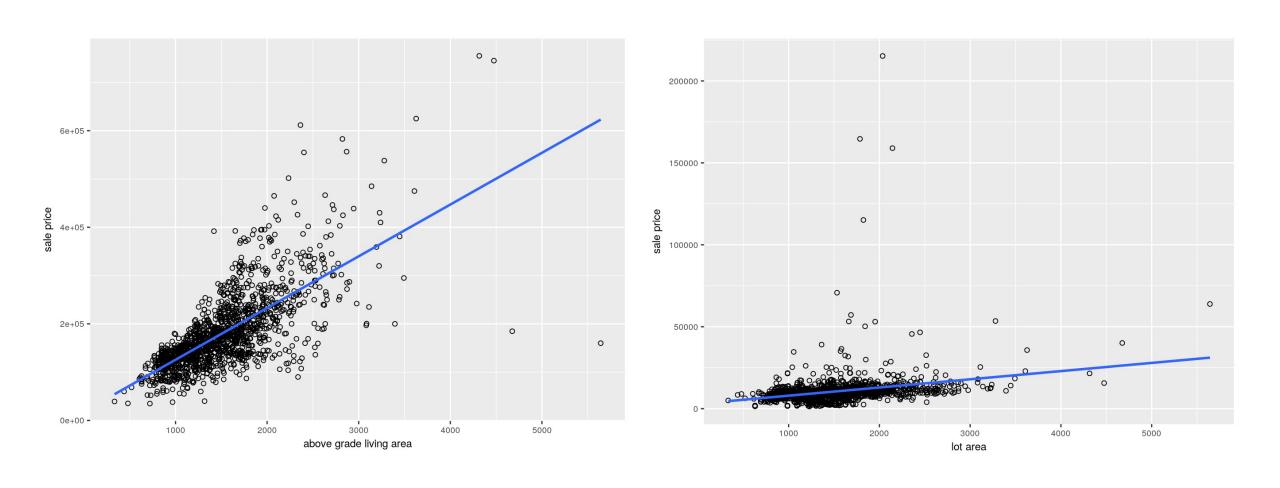
房屋交易資料

	price	bedrooms	bathrooms	sqft_living	sqft_lot	waterfront	sqft_above	sqft_basement	yr_built	yr_renovated	zipcode
0	221900.0	3	1.00	1180	5650	0	1180	0	1955	0	98178
1	538000.0	3	2.25	2570	7242	0	2170	400	1951	1991	98125
2	180000.0	2	1.00	770	10000	0	770	0	1933	0	98028
3	604000.0	4	3.00	1960	5000	0	1050	910	1965	0	98136
4	510000.0	3	2.00	1680	8080	0	1680	0	1987	0	98074

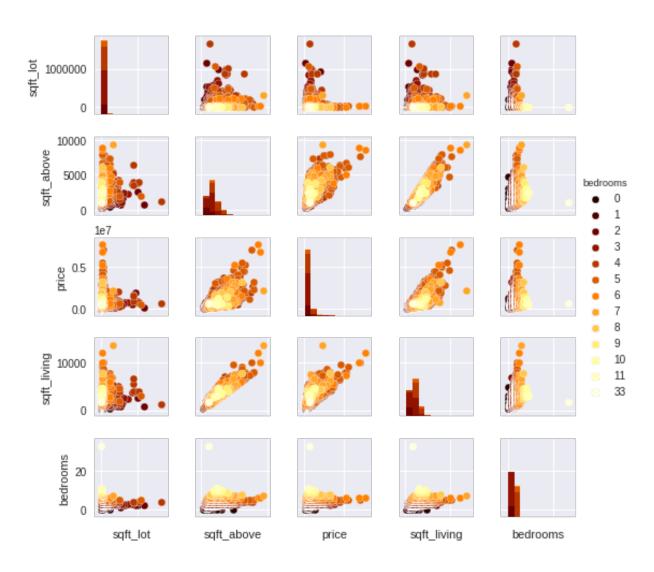
房屋屬性

id	a notation for a house	Numeric
year	date house was sold	String
month	date house was sold	String
day	date house was sold	String
price	Price is prediction target	Numeric
bedrooms	Number of Bedrooms/House	Numeric
bathrooms	Number of bathrooms/bedrooms	Numeric
sqft_living	square footage of the home	Numeric
sqft_lot	square footage of the lot	Numeric
floors	Total floors (levels) in house	Numeric
waterfront	House which has a view to a waterfront	Numeric
view	Has been viewed	Numeric
condition	How good the condition is (Overall)	Numeric
grade	overall grade given to the housing unit	Numeric
sqft_above	square footage of house apart from basement	Numeric
sqft_basement	square footage of the basement	Numeric
yr_built	Built Year	Numeric
yr_renovated	Year when house was renovated	Numeric
zipcode	zip	Numeric
lat	Latitude coordinate	Numeric
long	Longitude coordinate	Numeric
sqft_living15	Living room area	Numeric
sqft_lot15	lotSize area	Numeric

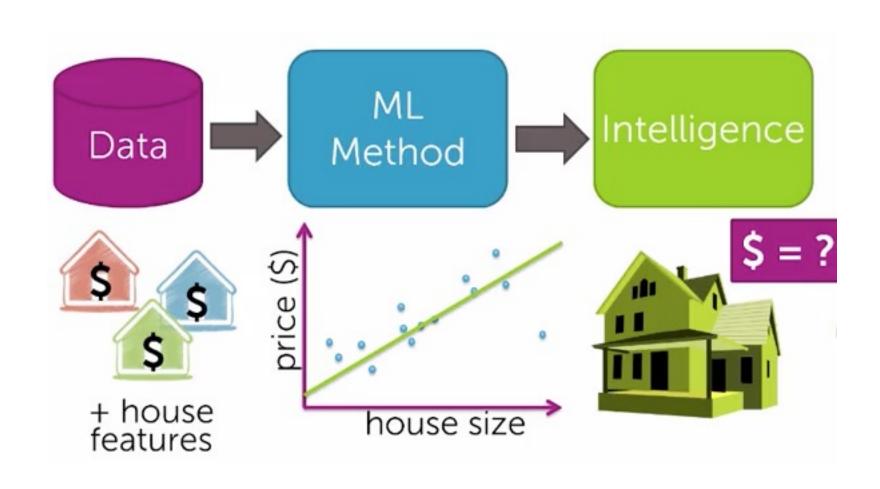
Correlation between a Feature and the Price



Correlation between Features

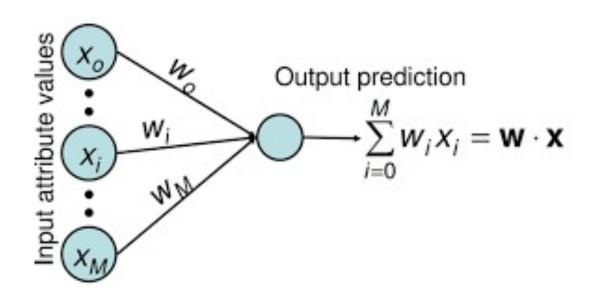


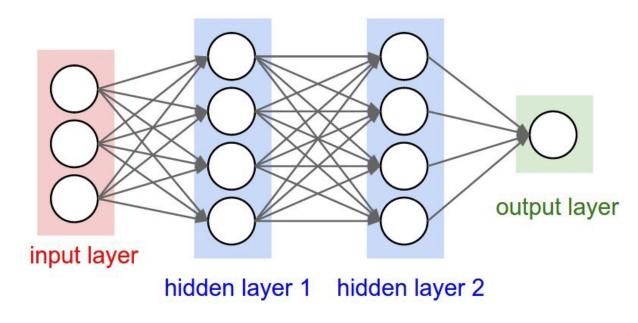
以機器學習演算法訓練回歸模型



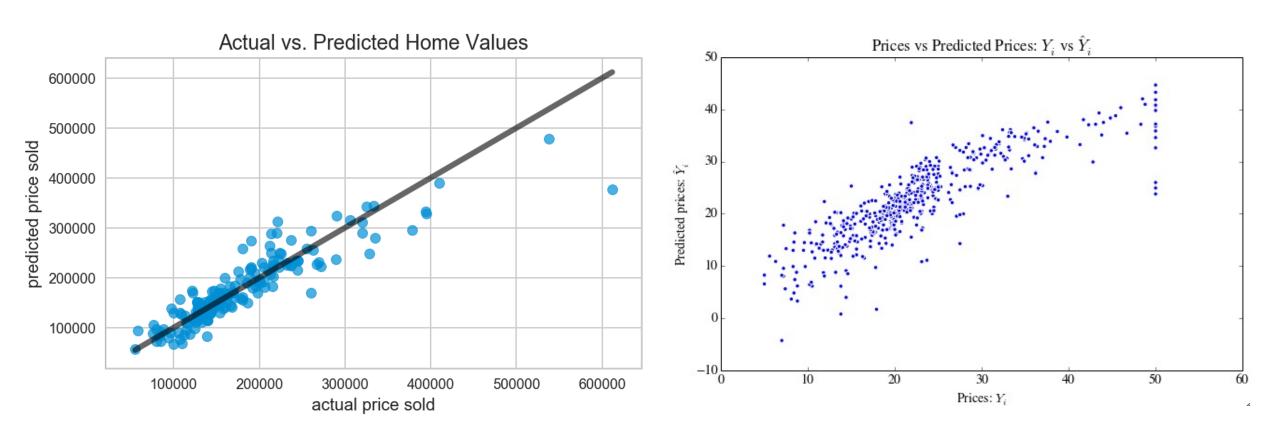
Neural Network-based Regression

• Linear vs. nonlinear model





Result Analysis





GitHub Classroom

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GitHub Education

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Your course assignments on GitHub

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Sign in with your GitHub account to get started







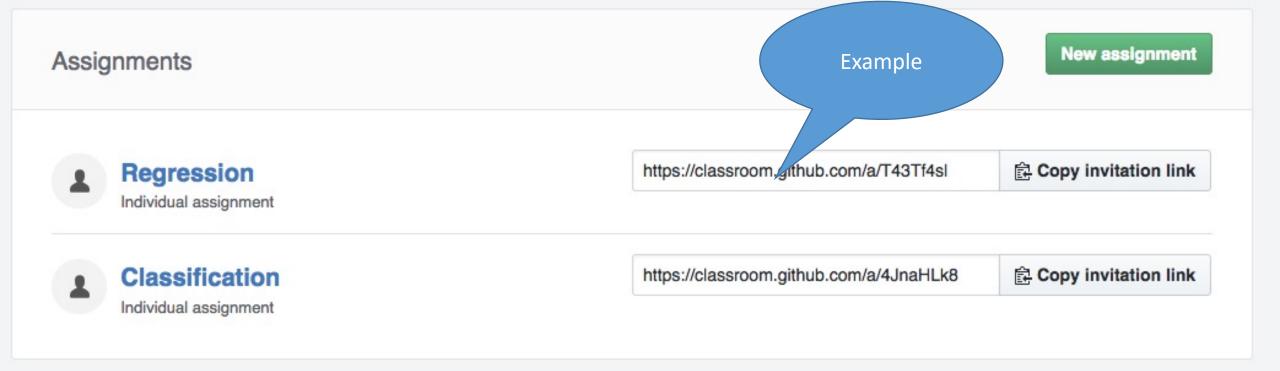




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MachineLearningNTUT

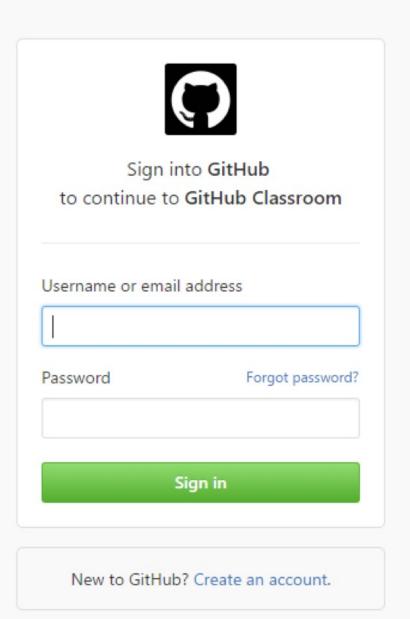
Manage classroom



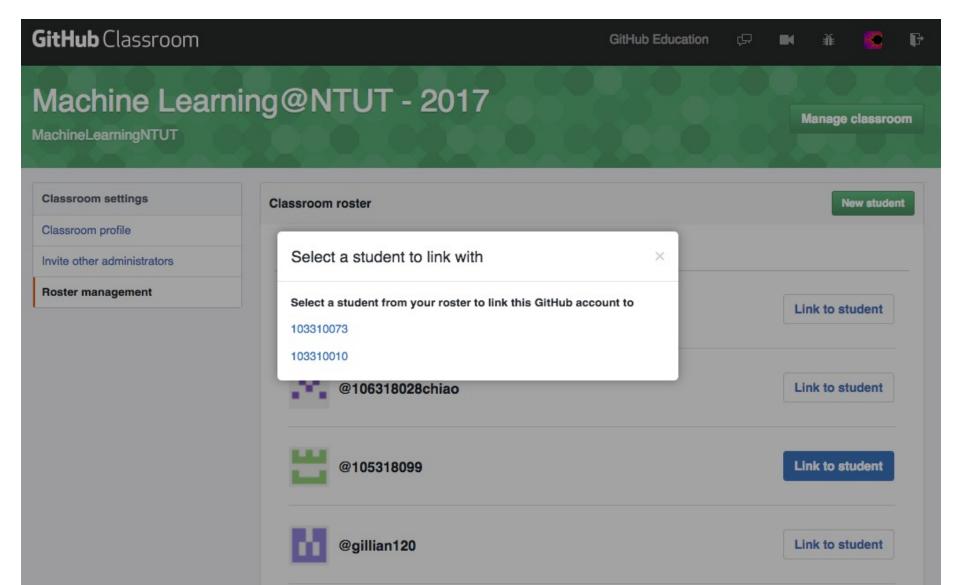
Create GitHub Classroom Account

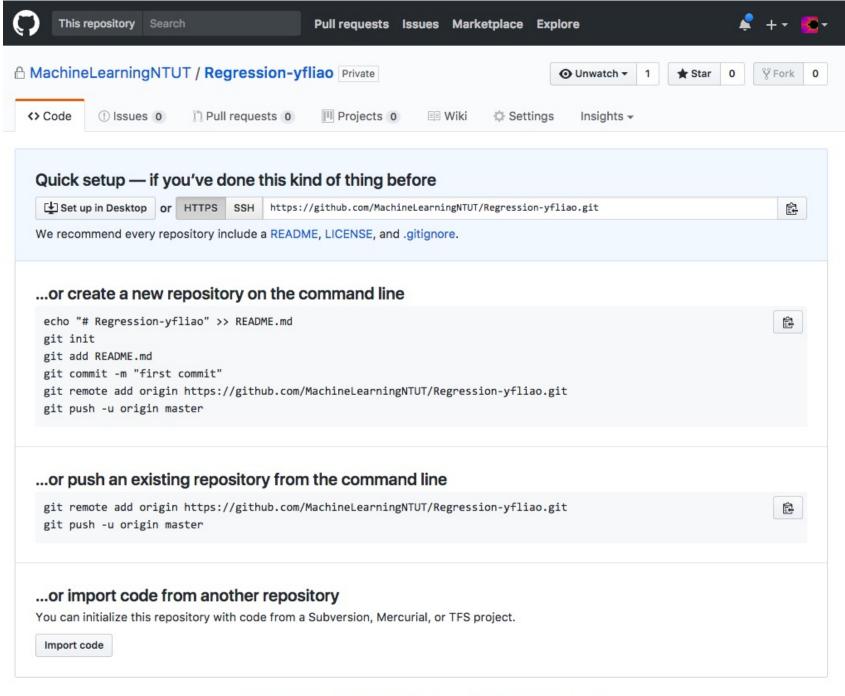
•以"學號+姓名"當你的帳號名稱



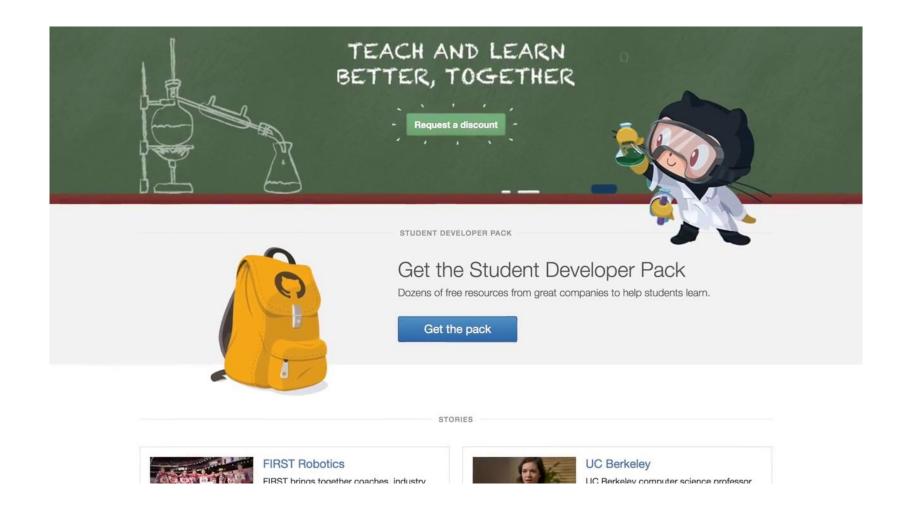


連結課程學生名單 (學號)





Student Developer Pack



Getting started with Kaggle

House Prices Competition



kaggle

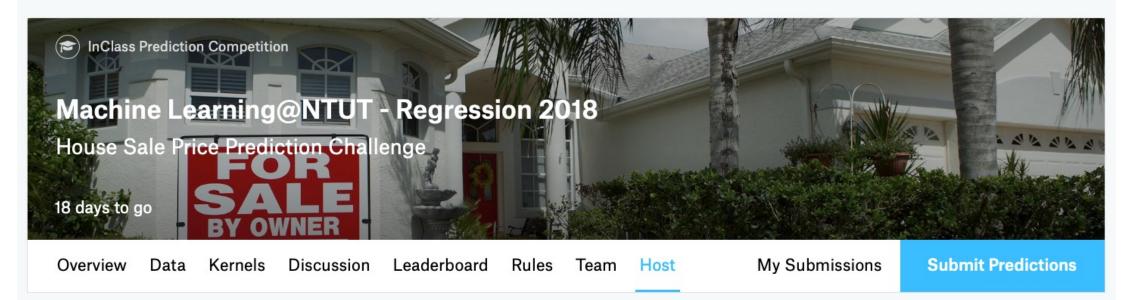
Search kaggle

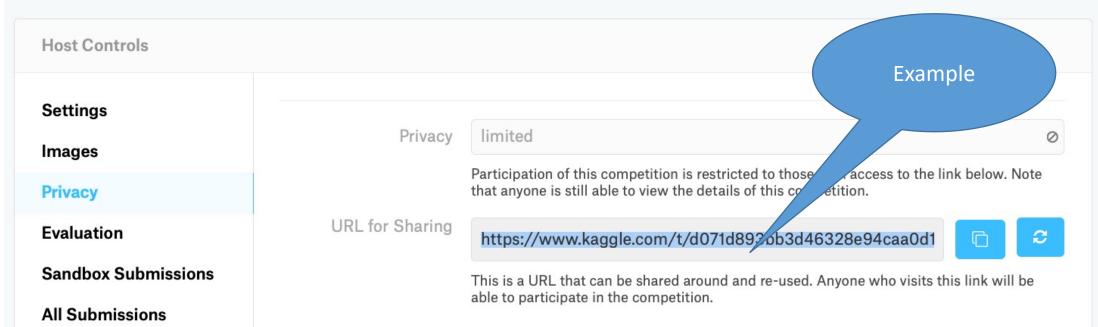
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Competitions

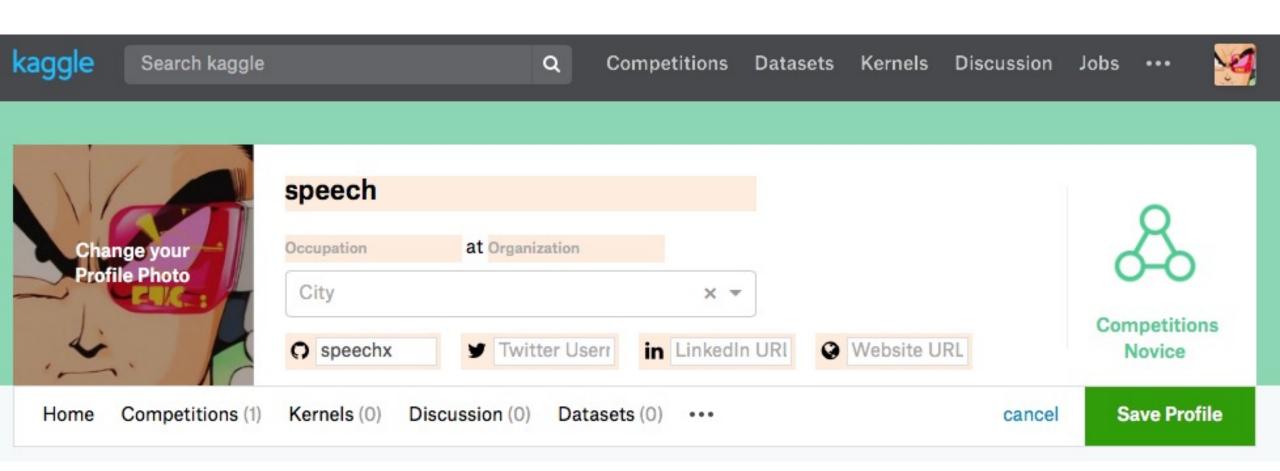
Datasets Kernels Discussion Learn







跟GitHub帳號連結





Overview

Edit

Description

Evaluation

+ Add Page

Please use the given house features to predict its sale price.

The given database for training a sale price predictor contain house sale prices for somewhere out there with 19 house features plus the price and the id columns.

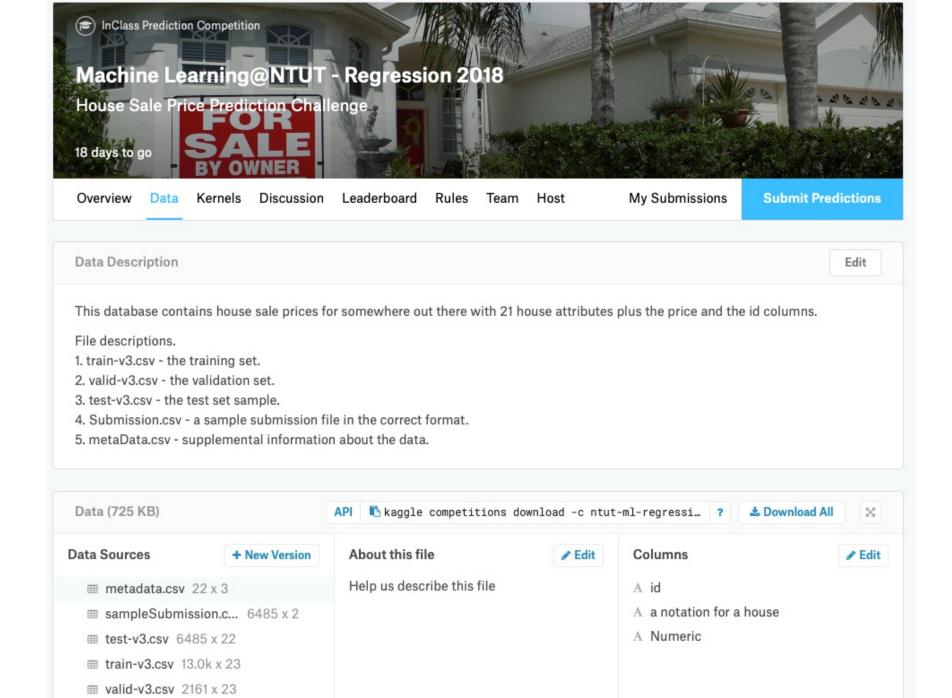
此次作業是使用回歸模型做房價預測,你們要做的是:

- 1. 用train.csv跟valid.csv訓練模型(一行是一筆房屋交易資料的紀錄,包括id, price與19種房屋參數)
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- 5. 嘗試改進預測模型

程式要放到Github Classroom

報告要包括:

- 1. 做法說明
- 2. 程式方塊圖與寫法
- 3. 畫圖做結果分析
- 4. 討論預測值誤差很大的,是怎麼回事?
- 5. 如何改進?





Overview Data Kernels Discussion Leaderboard Rules Team Host My Submissions **Submit Predictions** Make a submission for yfliao You have 10 submissions remaining today. This resets 18 hours from now (00: 00 UTC). Step 1 Upload submission file **Upload Submission File** Number of Predictions File Format We expect the solution file to have 6485 prediction rows. This file Your submission should be in CSV format. You can upload this in a zip/gz/rar/7z should have a header row. Please see sample submission file on archive, if you prefer. the data page. M♣ Styling with Markdown supported Step 2 Describe submission

Make Submission

Example Codes

- Tensorflow
 - https://www.tensorflow.org/tutorials/keras/regression
- Keras
 - https://github.com/sunyam/HousePrices_Regression
- Kaggle
 - https://www.kaggle.com/c/house-prices-advanced-regression-techniques
- References
 - Regression Tutorial with the Keras Deep Learning Library in Python
 - http://machinelearningmastery.com/regression-tutorial-keras-deep-learning-library-python/
 - Model Evaluation and Validation: Predicting Boston Housing Prices
 - https://olegleyz.github.io/boston_housing.html

Example: Boston House Price Dataset

506

No

ta Set

Characteristics:

Multivariate

Number of Instances:

Area:

N/A

Attribute

Categorical, Integer,

Number of Attributes:

Date Donated

1993-07-07

Characteristics:

Associated Tasks:

Real

Regression

Missing Values?

Number of Web Hits:

328614

Attribute Information:

- 1. CRIM: per capita crime rate by town
- 2. ZN: proportion of residential land zoned for lots over 25,000 sq.ft.
- 3. INDUS: proportion of non-retail business acres per town
- 4. CHAS: Charles River dummy variable (= 1 if tract bounds river; 0 otherwise)
- 5. NOX: nitric oxides concentration (parts per 10 million)
- 6. RM: average number of rooms per dwelling
- 7. AGE: proportion of owner-occupied units built prior to 1940
- 8. DIS: weighted distances to five Boston employment centres
- 9. RAD: index of accessibility to radial highways
- 10. TAX: full-value property-tax rate per \$10,000
- 11. PTRATIO: pupil-teacher ratio by town
- 12. B: 1000(Bk 0.63)² where Bk is the proportion of blacks by town
- 13. LSTAT: % lower status of the population
- 14. MEDV: Median value of owner-occupied homes in \$1000's



• Download: Boston house price dataset