## Data science HW2

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Date: March 29<sup>th</sup>, 2022

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### HW2

- Description
- How to submit and choose predictions
- Baseline method
- Hints

### Kaggle

- HW2 will be held on Kaggle
  - Please register a Kaggle account first
- A platform of
  - Machine learning competition
  - Sharing dataset
- https://zh.wikipedia.org/wiki/Kaggle



#### HW2

#### **NTHU DS2022 HW2**

NTHU data science 2022 spring HW2

- HW2 Kaggle link
  - https://www.kaggle.com/t/5e802ad74032485e94a4e1d1d91c0870
- Deadline: 2021/04/19 23:59 (3 weeks)
- We will use the result on Kaggle to score this homework
  - Please hand in a python file
  - Remember to fill your Kaggle name in the google form https://docs.google.com/spreadsheets/d/1nh0shU11SOPmwZFAzIVv71yB2olc IMHW/edit?usp=sharing&ouid=108250933224256718627&rtpof=true&sd=true

### Problem description

- Supervised binary classification problem
- Given a data set
  - Training set with label
  - Testing set without
- You need to predict the labels of testing data

### Dataset description

- The dataset is **transformed** from real weather observations dataset
- 16 numeric features, 5 nominal features, 1 label
  - Numeric feature are nonlinear transformed
  - About 20% data become missing value

Our dataset label is 'Label'

### Output format

- For each testing instance, there is a unique id
- Output your prediction to csv file with the following format and submit to kaggle

#### Remember to output the first line

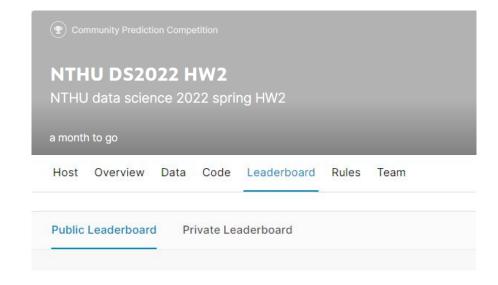
- Id, Label
- Id1, Label 1
- Id2, Label 2

• ...

Id		Label	
	0		(
	1		(
			(
	2		(
	4		(
	5		(
	6		(

#### **Evaluation**

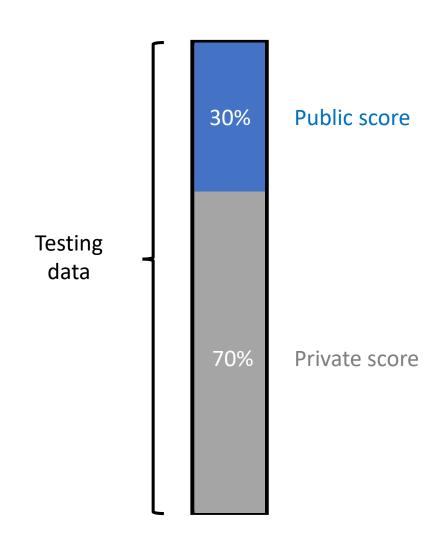
- We use F1-score
  - $2 \times \frac{precision \times recall}{precision + recall}$
- There are two leaderboards on Kaggle
  - Public
    - Can be seen during competition
  - Private
    - Can be seen after competition



#### Public and Private leaderboard

- Public (Can be seen during competition)
  - 30% testing data
  - For reference

- **Private** (Can be seen after competition)
  - the other 70%
  - Use this result for final scoring

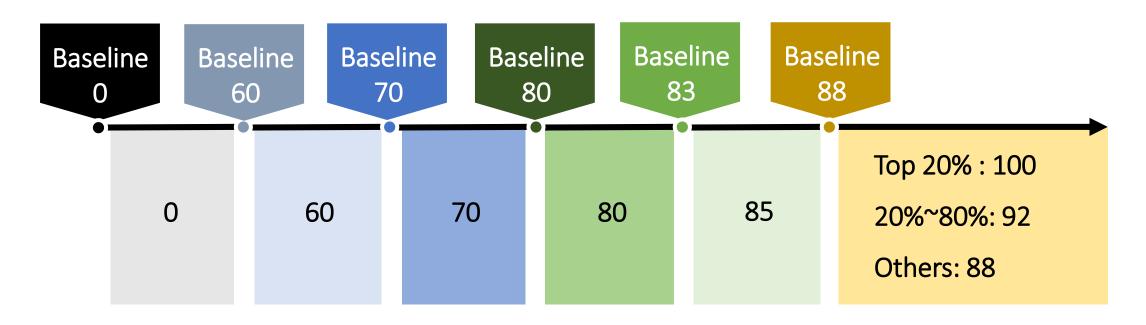


### Scoring

- Use private leaderboard result for final scoring
- Baseline scores
  - We will score according to given 7 baseline scores

	Public	Private
Baseline 88	0.44492	0.43355
Baseline 83	0.40148	0.38576
Baseline 80	0.36224	0.36259
Baseline 70	0.32343	0.33192
Baseline 60	0.29102	0.28034
Baseline 0	0.26840	0.25701

### Scoring



- You will get **0**, if your private score is between baseline 0 and baseline 60
- You will get **60**, if your private score is between baseline 60 and baseline 70
- You will get **70**, if your private score is between baseline 70 and baseline 80
- And so on

## Scoring

- Baseline scores
  - There are benchmarks on the leaderboard for reference

#	Team	Members	Score	Entries	Last	Code
<b>/</b>	Baseline 88		0.44492			
<b>/</b>	Baseline 83		0.40411			
<b>/</b>	Baseline 80		0.36244			
<b>/</b>	Baseline 70		0.32343			
<b>/</b>	Baseline 60		0.29102			
۴ï	Baseline 0		0.26804			
۴Ï	sampleSubmission.csv		0.15929			

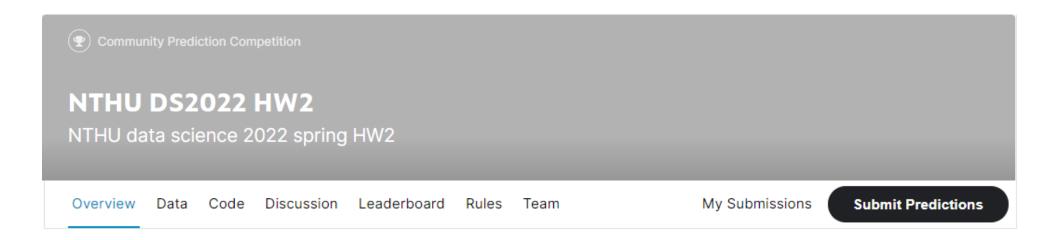
#### Other rules

- You can submit 15 times per day
- You can choose 4 predictions for final scoring
  - Kaggle will use the best one to be your final result

How to submit and choose predictions

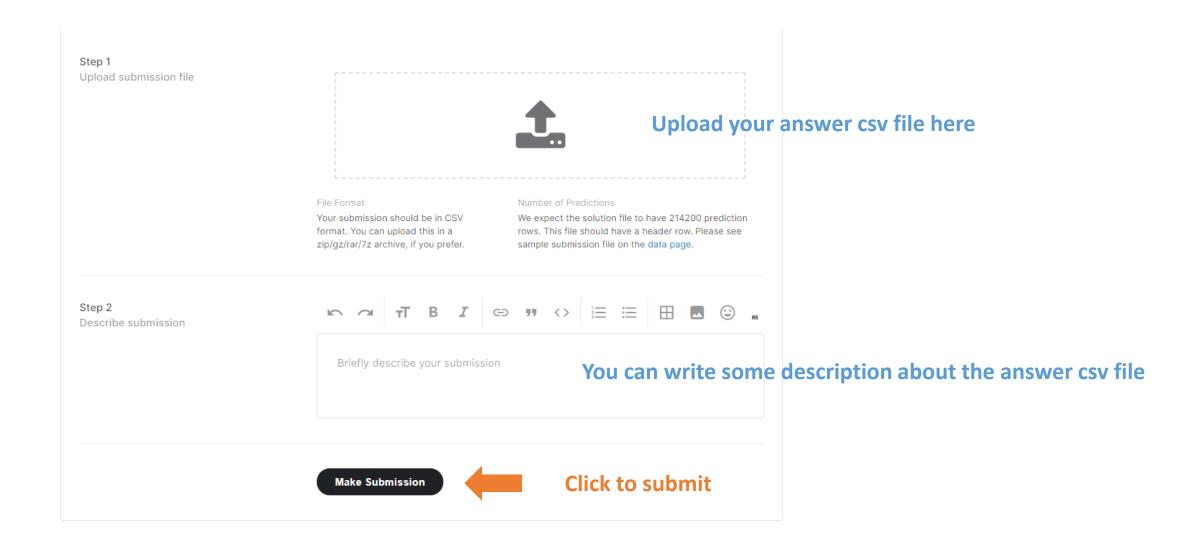
#### How to submit

• Click 'Submit Predictions' button on the navigation bar



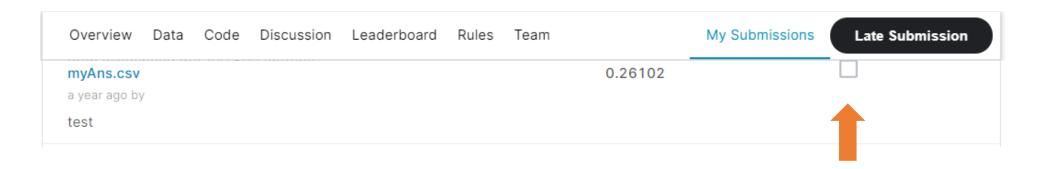


### How to submit



### Choose predictions for final scoring

You can see all your submissions in 'My Submissions'



Remember to choose 4 predictions before the deadline

- We provide a simple baseline method code for your reference
  - Baseline 0
- The steps in baseline are as below
  - Read training/testing data
  - Drop columns which are not numeric features
  - Fill missing value
  - Train a decision tree classifier
  - Output prediction

Read training/testing data

```
# 為了處理方便,把 'train.csv' 和 'test.csv' 合併起來,'test.csv'的 Weather 欄位用 0 補起來。

df = pd.read_csv('train.csv')

df_test = pd.read_csv('test.csv')

df_test['Label'] = np.zeros((len(df_test),))

# 以 train_end_idx 作為 'train.csv' 和 'test.csv' 分界列,

train_end_idx = len(df)

df = pd.concat([df, df_test], sort=False)
```

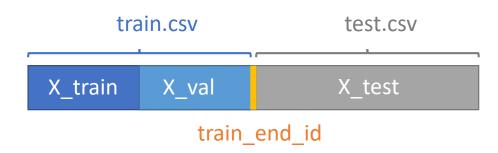
- Drop columns which are not numeric features
- Fill missing value

```
# 將非數值欄位拿掉

df = df.drop(columns = [col for col in df.columns if df[col].dtype == np.object])

# 將 missing value 補 0

df = df.fillna(0)
```



Split dataset

```
from sklearn.model_selection import train_test_split

X_train, X_val, y_train, y_val = train_test_split(
    df.drop(columns = ['Label']).values[:train_end_idx, :],
    df['Label'].values[:train_end_idx], test_size=0.5)

X_test = df.drop(columns = ['Label']).values[train_end_idx:, :]
```

Train a decision tree classifier and output prediction

```
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy score, f1 score
#train tree model
model = DecisionTreeClassifier()
model.fit(X train,y train)
#predict
y pred decision = model.predict(X val)
print('Accuracy: %f' % accuracy score(y val, y pred decision))
print('f1-score: %f' % f1 score(y val, y pred decision))
ans pred = model.predict(X test)
df sap = pd.DataFrame(ans pred.astype(int), columns = ['Label'])
df sap.to csv('myAns.csv', index label = 'Id')
```

- You can try to encode features in object type
  - Some features in object type may contain important information

```
from sklearn.preprocessing import LabelEncoder
labelencoder = LabelEncoder()
df['Loc'] = labelencoder.fit_transform(df['Loc'])
...
```

Fillna with median in numeric features instead of 0

```
df[i] = df[i].fillna(median)
```

Complete these may achieve the same or higher effect as the baseline 60

- Try different models
  - KNN, SVM, Logistic Regression, Random Forest ...

```
from sklearn.neighbors import KNeighborsClassifier
from sklearn.svm import SVC
from sklearn.linear_model import LogisticRegression
from sklearn.ensemble import RandomForestClassifier
from sklearn.naive_bayes import GaussianNB
```

Finetune the model may achieve higher effect than the baseline 70

- The numbers of label is imbalance
- Deal with data imbalance
  - There are some sampler you can try

from imblearn.over\_sampling import SMOTE, ADASYN, RandomOverSampler

Complete this may achieve the same or higher effect as the baseline 80

- More techniques for better performance
  - Feature selection
  - Normalization
  - Dimension reduction (PCA, TSNE)
  - Try other different models (AdaBoost....)
  - ...
- We use private leaderboard as the final score
  - Use public score to choose your model is dangerous
  - It's better to perform validation

### Packages you may use

- Scikit-learn
  - https://scikit-learn.org/stable/index.html
- Pandas
  - https://pandas.pydata.org/pandas-docs/stable/
- Imbalance learn (for over sampling and down sampling)
  - https://imbalanced-learn.readthedocs.io/en/stable/