→ Lab 5: Kaggle Competition

▼ Notebook by: Davis Ulrich

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```
import pandas as pd
from sklearn.neural_network import MLPClassifier
from sklearn.cluster import KMeans
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import silhouette score
from sklearn.model selection import KFold
from sklearn.preprocessing import StandardScaler, MinMaxScaler
from sklearn.preprocessing import LabelEncoder, OneHotEncoder
from sklearn.feature_extraction import DictVectorizer
from sklearn.pipeline import Pipeline
from sklearn.metrics import accuracy_score
from sklearn.model_selection import train_test_split
import numpy as np
import matplotlib.pyplot as plt
train = pd.read_csv('/content/drive/My Drive/Data 144/train.csv')
test = pd.read_csv("/content/drive/My Drive/Data 144/test.csv")
train
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	C
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	٤
				Allen Mr								

test

	Saved successfully!		Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
Saved successf			elly, Mr. James	male	34.5	0	0	330911	7.8292	NaN	Q
1	893	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	363272	7.0000	NaN	S
2	894	2	Myles, Mr. Thomas Francis	male	62.0	0	0	240276	9.6875	NaN	Q
3	895	3	Wirz, Mr. Albert	male	27.0	0	0	315154	8.6625	NaN	S
4	896	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	3101298	12.2875	NaN	S

Dealing with Nans

Dona. Fermina

```
train.isna().sum()
    PassengerId
    Survived
    Pclass
    Name
    Sex
                 177
    Age
    SibSp
    Parch
    Ticket
    Fare
                  687
    Cabin
    Embarked
    dtype: int64
test.isna().sum()
    PassengerId
    Pclass
    Name
    Sex
    Age
                  86
    SibSp
    Parch
    Ticket
    Fare
                   1
                  327
    Cabin
    Embarked
    dtype: int64
# The Fare feature is a weird object that you can't average or anyting
train['Fare'] = [float(i) for i in train['Fare']]
test['Fare'] = [float(i) for i in test['Fare']]
# Many more people who have a missing age did not survive
train.sort_values('Age').iloc[800:]['Survived'].mean()
    0.21978021978021978
# Since most passengers missing their age didn't survive,
# it's relevant to keep those columns and be able to distinguish them
train['Age'] = train[['Age']].fillna(999.0)
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarke
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NA	
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NA	
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	
				Allen Mr								

→ Adding Features / One-Hot Encoding

```
train['Sex'] = [1 if sex == 'male' else 0 for sex in train['Sex']]
test['Sex'] = [1 if sex == 'male' else 0 for sex in test['Sex']]
train
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	1	22.0	1	0	A/5 21171	7.2500	NA	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	0	38.0	1	0	PC 17599	71.2833	C85	С
2	3	1	3	Heikkinen, Miss. Laina	0	26.0	0	0	STON/O2. 3101282	7.9250	NA	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	0	35.0	1	0	113803	53.1000	C123	S
				Allen Mr								

```
# Out of those that have long names (greater than 35 characters), 71% survived.

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# train['long_name'] = [int(len(name) >= 35) for name in train['Name']]

# test['long_name'] = [int(len(name) >= 35) for name in test['Name']]

train
```

0.7071428571428572

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	1	22.0	1	0	A/5 21171	7.2500	NA	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	0	38.0	1	0	PC 17599	71.2833	C85	С
2	3	1	3	Heikkinen, Miss. Laina	0	26.0	0	0	STON/O2. 3101282	7.9250	NA	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	0	35.0	1	0	113803	53.1000	C123	S
				Allen Mr								

```
# Length of name includes long and short
train['name_len'] = [len(name) for name in train['Name']]
test['name_len'] = [len(name) for name in test['Name']]
```

```
# Some embarking locations have higher average survival rate than others
print(train[train['Embarked'] == 'S']['Survived'].mean())
```

```
train = train.merge(pd.get_dummies(train['Embarked']), left_index = True, right_index = True)
test = test.merge(pd.get_dummies(test['Embarked']), left_index = True, right_index = True)
train
```

```
0 336056521730130//3
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             891.000000
    count
              32.204208
    mean
    std
              49.693429
    min
               0.000000
    25%
               7.910400
    50%
              14.454200
    75%
              31.000000
             512.329200
    max
    Name: Fare, dtype: float64
data = [train, test]
titles = {"Mr": 1, "Miss": 2, "Mrs": 3, "Master": 4, "Not Common": 5}
for dataset in data:
    # extract titles
    dataset['Title'] = dataset.Name.str.extract(' ([A-Za-z]+)\.', expand=False)
    dataset['Title'] = dataset['Title'].replace('Mme', 'Mrs')
    dataset['Title'] = dataset['Title'].replace('Mlle', 'Miss')
    dataset['Title'] = dataset['Title'].replace('Ms', 'Miss')
   dataset['Title'] = dataset['Title'].replace(['Lady', 'Countess','Capt', 'Col','Don', 'Dr','Major', 'Rev'
    # convert titles into numbers
    dataset['Title'] = dataset['Title'].map(titles)
    dataset['Title'] = dataset['Title'].fillna(0)
train = train.drop(['Name'], axis=1)
test = test.drop(['Name'], axis=1)
data = [train, test]
for dataset in data:
    dataset['number relatives'] = dataset['SibSp'] + dataset['Parch']
data = [train, test]
for dataset in data:
    dataset.loc[dataset['number relatives'] > 0, 'Alone?'] = 0
    dataset.loc[dataset['number relatives'] == 0, 'Alone?'] = 1
    dataset['Alone?'] = dataset['Alone?'].astype(int)
```

▼ Train Validation Split

```
# Train Val Split
X_train, X_val, Y_train, Y_val = train_test_split(train.drop(['Survived'], axis = 1), train['Survived'], ran
```

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Models

▼ Model 1: Decision Tree

```
train.columns
    Index(['PassengerId', 'Survived', 'Pclass', 'Sex', 'Age', 'SibSp', 'Parch',
            'Ticket', 'Fare', 'Cabin', 'Embarked', 'name_len', 'C', 'NA', 'Q', 'S',
            'Title', 'number relatives', 'Alone?'],
           dtype='object')
all_features = ['Pclass', 'Sex', 'Age', 'SibSp', 'Parch', 'Ticket', 'Fare', 'Cabin', 'Embarked', 'name_len',
       'Title', 'number relatives', 'Alone?']
features = ['Pclass', 'Sex', 'Age', 'SibSp', 'Parch', 'Fare', 'name_len', 'C', 'Q', 'S', 'NA', 'Title', 'num
tree = DecisionTreeClassifier(random_state = 42)
tree.fit(X_train[features], Y_train)
y pred tree = tree.predict(X val[features])
accuracy_score(Y_val, y_pred_tree)
    0.7802690582959642
# Best so far
# 73.5 ['Pclass', 'Sex', 'Age', 'Fare']
# 78.5 ['Sex']
# 80.7 ['Pclass', 'Sex', 'Age', 'SibSp', 'Parch', 'Fare', 'long_name', 'name_len', 'C', 'Q', 'S']
all_features = ['Pclass', 'Sex', 'Age', 'SibSp', 'Parch', 'Fare', 'name_len', 'C', 'Q', 'S', 'NA']
featuresm = ['Sex', 'Age', 'Fare']
tree2 = DecisionTreeClassifier(max depth=1, random state = 42)
tree2.fit(X_train[featuresm], Y_train)
y_pred_tree2 = tree2.predict(X_val[featuresm])
accuracy_score(Y_val, y_pred_tree2)
    0.7847533632286996
```

▼ Model 3: Ensemble Learning

```
# Taking the majority vote from the two decision trees and the neural network
ensemble = y_pred_tree + y_pred_tree2 + y_pred_NN
ensemble = np.round(ensemble/3, 0)
ensemble
□→ array([1., 0., 0., 1., 1., 1., 1., 0., 1., 1., 0., 0., 0., 0., 0., 1., 0.,
           1., 0., 0., 0., 0., 1., 0., 0., 0., 0., 0., 0., 1., 0., 1., 0., 1.,
           0., 0., 0., 1., 1., 0., 0., 0., 0., 1., 0., 0., 0., 0., 0., 1.,
           1., 0., 0., 0., 1., 0., 1., 1., 1., 0., 1., 1., 0., 0., 1., 0., 0.,
           0., 1., 1., 1., 0., 1., 0., 0., 1., 1., 1., 1., 0., 1., 1., 0., 0.,
           0., 1., 1., 0., 0., 1., 1., 0., 0., 0., 0., 0., 0., 1., 0., 0.,
           1., 0., 0., 0., 1., 0., 0., 1., 0., 1., 0., 1., 0., 0., 0., 1.,
           1., 0., 1., 1., 0., 0., 1., 1., 0., 1., 0., 1., 0., 0., 1., 1., 1.,
           1., 0., 0., 1., 0., 1., 0., 0., 1., 1., 0., 0., 1., 0., 0., 0., 0.,
           0., 0., 0., 0., 1., 1., 1., 0., 1., 0., 1., 0., 0., 0., 1., 0., 0.,
           1., 1., 0., 1., 0., 0., 1., 1., 1., 0., 0., 0., 0., 1., 1., 0., 0.,
           0., 0., 1., 0., 0., 0., 0., 1., 1., 1., 0., 1., 0., 0., 1., 1., 0.,
           0., 1., 0., 0., 0., 0., 0., 1., 0., 0., 0., 1., 0., 1., 0., 0.,
           1., 0.])
accuracy score(Y val, ensemble)
    0.8071748878923767
```

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▼ Testing

test

	PassengerId	Pclass	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked	name_len	С	Q
0	892	3	1	34.5	0	0	330911	7.8292	NA	Q	16	0	1
1	893	3	0	47.0	1	0	363272	7.0000	NA	S	32	0	0
2	894	2	1	62.0	0	0	240276	9.6875	NA	Q	25	0	1
3	895	3	1	27.0	0	0	315154	8.6625	NA	S	16	0	0
4	896	3	0	22.0	1	1	3101298	12.2875	NA	S	44	0	0
413	1305	3	1	999.0	0	0	A.5. 3236	8.0500	NA	S	18	0	0
414	1306	1	0	39.0	0	0	PC 17758	108.9000	C105	С	28	1	0
415	1307	3	1	38.5	0	0	SOTON/O.Q. 3101262	7.2500	NA	S	28	0	0
416	1308	3	1	999.0	0	0	359309	8.0500	NA	S	19	0	0
417	1309	3	1	999.0	1	1	2668	22.3583	NA	С	24	1	0

test[features]

	Pclass	Sex	Age	SibSp	Parch	Fare	name_len	С	Q	s
0	3	1	34.5	0	0	7.8292	16	0	1	0
1	3	0	47.0	1	0	7.0000	32	0	0	1
2	2	1	62.0	0	0	9.6875	25	0	1	0
3	3	1	27.0	0	0	8.6625	16	0	0	1
4	3	0	22.0	1	1	12.2875	44	0	0	1
413	3	1	999.0	0	0	8.0500	18	0	0	1
414	1	0	39.0	0	0	108.9000	28	1	0	0
415	3	1	38.5	0	0	7.2500	28	0	0	1
416	3	1	999.0	0	0	8.0500	19	0	0	1
417	3	1	999.0	1	1	22.3583	24	1	0	0

418 rows × 10 columns

NN.predict(test[features])

```
array([0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 1,
                               0, 0, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1,
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                               0, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 1,
                               1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1,
          1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0,
          0, 1, 1, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1,
          0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1,
          0, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1,
          0, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1,
          0, 1, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 0, 1, 0,
          1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1,
          1, 0, 0, 0, 1, 0, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1,
          1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 1, 0,
          0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1,
          0, 1, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0,
          1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1, 0,
          0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0,
          1, 1, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1,
          0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 0]
```

submission = pd.DataFrame(index=test.PassengerId)
submission['Survived'] = NN.predict(test[features])
submission['Survived'] = tree2.predict(test[featuresm])
submission

Survived

PassengerId									
892	0								
893	1								
894	0								
895	0								
896	1								
1305	0								
1306	1								
1307	0								
1308	0								
1309	0								

418 rows x 1 columns

submission.reset_index().to_csv('submission.csv', index=False)

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