

Titanics Dataset Predictions

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Mounted Google Drive to import data from Kaggle Train, and Test data sets.

```
In [1]: from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

```
In [2]: import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
from scipy.stats import reciprocal
import tensorflow as tf
import tensorflow.keras as keras
from tensorflow.keras.callbacks import *
from tensorflow.keras.layers import *
from tensorflow.keras.metrics import *
from tensorflow.keras.models import *
from tensorflow.keras.optimizers import *
from tensorflow.keras.optimizers.schedules import *
import tensorflow.keras.callbacks as tkc
from tensorflow.keras.wrappers.scikit_learn import KerasClassifier
from sklearn.compose import make_column_transformer, ColumnTransformer
from sklearn.impute import SimpleImputer
from sklearn.manifold import Isomap
from sklearn.model_selection import cross_val_score, GridSearchCV
from sklearn.metrics import mean_squared_error
from sklearn.pipeline import Pipeline
from sklearn.preprocessing import StandardScaler, OneHotEncoder, OrdinalEncoder, LabelEncoder
from sklearn.utils import resample
```

Here I included standard packages which include TensorFlow machine learning library, as well as the Sci-Kit Learn library for data processing. As well as standard imports such as Pandas for reading data, and numpy for numerical analysis.

```
In [3]: file1_train, file2_test = pd.read_csv(r"/content/drive/MyDrive/Titanic/train.csv", delimiter=",", \
pd.read_csv(r"/content/drive/MyDrive/Titanic/test.csv", delimiter= ",")
```

Now I am going to encode the Sex column into a binary 1s, and 0s output Sex_Binary column.

```
In [4]: file1_train = file1_train.drop(columns = ["PassengerId", "Name"])
X, y = file1_train.drop("Survived", axis=1), file1_train["Survived"]
```

```
In [6]: file1_train.head()
```

```
Out[6]:
```

	Survived	Pclass	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	0	3	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	1	1	female	38.0	1	0	PC 17599	71.2833	C85	C
2	1	3	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	1	1	female	35.0	1	0	113803	53.1000	C123	S
4	0	3	male	35.0	0	0	373450	8.0500	NaN	S

```
In [7]: objs = X.select_dtypes(["object"])
num = X.select_dtypes(["number"])
```

```
In [8]: objs.isnull().sum().head()
```

```
Out[8]: Sex          0
Ticket          0
Cabin         687
Embarked        2
dtype: int64
```

```
In [9]: num.isnull().sum().head()
```

```
Out[9]: Pclass        0
Age          177
```

```
SibSp      0
Parch      0
Fare       0
dtype: int64
```

```
In [10]: numerical_features = num.columns
numerical_pipeline = Pipeline(
    steps=[
        ("imputer", SimpleImputer(strategy = 'mean')),
        ("scaler", StandardScaler())
    ])
```

```
In [11]: for i in range(len(objs.columns)):
        print(objs.columns[i], objs.iloc[:,i].value_counts().unique())
```

```
Sex [577 314]
Ticket [7 6 5 4 3 2 1]
Cabin [4 3 2 1]
Embarked [644 168 77]
```

```
In [12]: binary = ["Sex"]
binary_pipeline = Pipeline(steps=[("binary", OneHotEncoder())])
```

```
In [13]: cat1 = ["Ticket", "Cabin", "Embarked"]
catergorical_pipeline = Pipeline(steps=[("imputer", SimpleImputer(strategy = 'most_frequent')), ("ordinal_en
```

```
In [14]: data_preprocessor = ColumnTransformer( [('numerical', numerical_pipeline, numerical_features),
        ('binary', binary_pipeline, binary),
        ('categorical', catergorical_pipeline, cat1)])
```

```
In [15]: X.head()
```

```
Out[15]:
```

	Pclass	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	3	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	1	female	38.0	1	0	PC 17599	71.2833	C85	C

	Pclass	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
2	3	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	1	female	35.0	1	0	113803	53.1000	C123	S

```
In [16]: X = Pipeline(steps=[('processing',data_preprocessor)]).fit_transform(X)
```

```
In [17]: param_distribs = {'learn_rate' : np.array(np.linspace(.1,.9))}
```

```
In [18]: def base_model1(learn_rate = .1):
    input_dim = X.shape[1]
    model =Sequential([
        Dense(200 , input_dim = input_dim, activation= "relu"),
        Dropout(rate=.10),
        Dense(100, activation= "tanh"),
        Dense(1,activation = "sigmoid"),
    ])
    lr_schedule = ExponentialDecay(
        learn_rate,
        decay_steps=100000,
        decay_rate=0.96,
        staircase=True)
    model.compile(loss='binary_crossentropy',optimizer=tf.keras.optimizers.Adam(learning_rate=lr_schedule),
    return model
```

```
In [19]: checkpoint = [ModelCheckpoint("Titanic.h5", monitor='accuracy', verbose=1, save_best_only=True, save_weights_
early = EarlyStopping(monitor='accuracy', min_delta=0, patience=10, verbose=1, mode='auto')]
```

```
In [21]: NN_clf = KerasClassifier(build_fn=base_model1, epochs=100, verbose=1, callbacks =[checkpoint,early] )
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:1: DeprecationWarning: KerasClassifier is deprecated, use Sci-Keras (<https://github.com/adriangb/scikeras>) instead.

"""Entry point for launching an IPython kernel.

```
In [23]: %%capture
random_trainor = GridSearchCV(estimator=NN_clf,param_grid=param_distribs, cv=None)
random_trainor.fit(X,(y.values.reshape(-1,1)))
```

```
In [24]: best = random_trainor.best_estimator_.model
```

```
In [25]: best.summary()
```

Model: "sequential_413"

Layer (type)	Output Shape	Param #
dense_1239 (Dense)	(None, 200)	2200
dropout_413 (Dropout)	(None, 200)	0
dense_1240 (Dense)	(None, 100)	20100
dense_1241 (Dense)	(None, 1)	101

=====
Total params: 22,401
Trainable params: 22,401
Non-trainable params: 0
=====

```
In [26]: best.save("Titanic.h5")
```

```
In [31]: X_test = data_preprocessor.fit_transform(file2_test)
```

```
In [32]: PassengerId = file2_test["PassengerId"].to_list()
```

```
In [33]: final_pred = (best.predict(X_test) > 0.5).astype("int32").flatten()
```

```
In [30]: #d = {"PassengerId":PassengerId, "Survived":final_pred}
#pd.DataFrame(data=d,index=None, columns= ["PassengerId","Survived"]).to_csv("12232021.csv",index=False, hea
```

```
In [ ]:
```

```
In [ ]:
```