Experiment No. 1: Linear regression using deep neural network on Boston housing dataset.

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from keras.callbacks import ModelCheckpoint
from keras.models import Sequential
from keras.layers import Dense, Activation, Flatten
from sklearn.model selection import train_test_split
from sklearn.ensenmble import RandomForestRegressor
from sklearn.metrics import mean_absolute_error
from matplotlib import pyplot as plt
import seaborn as sb
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
import warnings
warnings.filterwarnings('ignore')
warnings.flterwarnings('ignore', category=DeprecationWarning)
from xgboost import XGBRegressor
#processing of Dataset
def get_data):
#get train data
train_data_path ='train.csv'
train = pd.read csv(train _data_path)
#get test data
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test_data_path ='test.csv'
test = pd.read _csv(test_data_path)
return train, test
def get_combined_data ():
#reading train data
train, test = get_data()
target = train.SalePrice
train.drop (['SalePrice'],axis = 1, inplace = True)
combined=train.append(test)
combined.reset_index(inplace=True)
combined.drop([index', 'ld'], inplace=True, axis=1)
return combined, target
#Load train and test data into pandas DataFrames
train_data, test_data=getdata()
#Combine train and test data to process them together
combined, target = get_combined_data()
# define a function to get the columns that don't have any missing values
def get_cols_with_no_nans(df,col_type):
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Arguments:
df: The dataframe to process
col_type:
num: to only get numerical columns with no nans
no_num: to only get nun-numerical columns with no nans
all: to get any columns with no nans
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if (col_type =='num'):
predictors = df.select_dtypes(exclude=['object'])
elif (col_type =='no_num'):
predictors = df.select_dtypes(include=['object'])
elif (col_type =='all):
predictors = df
else
print('Error : choose a type (num, no_num, all)')
return 0
cols_with_no_nans = []
for col in predictors.columns:
if not df[col].isnull().any():
cols_with_no_nans.append(col)
return cols_with_no_nans
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return cols_with_no_nans

# Get the columns that do not have any missing values.

num_cols = get_cols_with_no_nans(combined, 'num')

cat_cols = get_cols_with_no_nans(combined, 'no_num')
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# Let's see how many columns we got
print (Number of numerical columns with no nan values:, len(num_cols))
print ('Number of nun-numerical columns with no nan values:, 'len(cat_cols))
[out]:
Number of numerical columns with no nan values: 25
Number of nun-numerical columns with no nan values: 20
#0ne Hot Encode The Categorical Features
def oneHotEncode(df,colNames):
for col in colNames:
if( df[col].dtype == np.dtype('object')):
dummies = pd.get_dummies(df[col],prefix=col)
df= pd.concat([df,dummies], axis=1)
#drop the encoded column
dfdrop([col],axis = 1, inplace=True)
return df
print("There were {} columns before encoding categorical
features'.format(combined.shape[1]))
combined = oneHotEncode(combined, cat_cols)
print('There are {} columns after encoding categorical features'.format(combined.shape[1]))
[out]:
There were 45 columns before encoding categorical features
There are 149 columns after encoding categorical features
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#split back combined dataFrame to training data and test data
def split combined ():
global combined
train = combined[:1460]
test = combined [1460:]
return train, test
train, test = split_combined()
#Making the Deep Neural Network
NN_model = Sequential()
# The Input Layer:
NN_model.add(Dense(128, kernel Initlalizer='normal', input_dim = train.shape [1],
activation='relu'))
# The Hidden Layers:
NN model.add(Dense(256, kernel Initiallzer='normal',activation='relu'))
NN model.add(Dense(256, kernel Initlalizer='normal',activation='relu'))
NN model.add(Dense(256, kernel_Initialízer='normal',activatlon='relu'))
# The Output Layer:
NN_ model.add (Dense(1, kernel_initializer='normal',activation='linear'))
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# Compile the network:
NN model.compile (loss='mean absolute error', optimlzer='adam', metrics=['mean
_absolute_error')
NN_model.summary()
[Out]:
-----Layer (type) Output Shape Param #
-----dense_1 (Dense) (None, 128) 19200
-----dense_2 (Dense) (None, 256) 33024
-----dense_3 (Derrse) (None, 256) 65792
-----dense_4 (Dense) (None, 256) 65792
-----dense_5 (Dense) (None, 1) 257
Total params: 184,065 Trainable params: 184,065 Non-trainable params: 0
#Define a checkpoint call back:
checkpoint_name = "Weights-{epoch:03d}--(val_loss:.5f).hdf5'
checkpoint = ModelCheckpoint (checkpoint_name, monitor='val_loss', verbose = 1,
save_best_only= True, mode ='auto')
callbacks_list = [checkpoint]
#Train the model:
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NN_model.fit(train, target, epochs=500, batch_size=32, validation_split = 0.2,
callbacks=callbacks_list)
[out]:
Train on 1168 samples, valildate on 292 samples
Epoch 1/500
1168/1168 [======]-0s 266us/step - loss: 19251.8903 - mean_absolute_error:
19251.8903 - val_loss: 23041.8968 - val_mean_absolute_error: 23041.8968
Epoch 00001: val loss did not improve from 21730.93555
Epoch 2/500
1168/1168 [======]- 0s 268us/step- loss: 18180.4985- mean _absolute_error:
18180,4985- val_loss: 22197.7991 - val _mean_absolute_error: 22197.7991
Epoch 00002: val_loss did not improve from 21730.93555
Epoch 00500: val loss did not improve from 18738.1983 1
# Load wights file of the best model:
wights_file = "Weights-478--18738.19831. hdf5' # choose the best checkpoint
NN_model.load_weights (wights_file) # load it
NN_model.compile(loss='mean absolute error'.
optimizer='adam',metrics=['mean absolute error'])
#Test the model
def make_submission (prediction, sub_name):
my_submission = pd.Data Frame({'ld':pd.read_csv('test.csv'). ld, 'SalePrice':prediction})
my_submission.to_csv('{}.csv'.format(sub_name), index=False)
print('A submission file has been made')
```

predictions = NN_model.predict(test)

OUT:

0.14605