Lab 1

Aim : Design and develop an Artificial Neural Network model which can predict the cost of insurance

import pandas as pd
import numpy as np
import seaborn as sb
import matplotlib.pyplot as plt

df = pd.read_csv("Insurance Costs.csv")

df

	person_id	age	sex	bmi	children	smoker	region	total_charges
0	1	19.0	female	27.900	0	yes	southwest	16884.9240
1	2	18.0	male	33.770	1	no	southeast	1725.5522
2	3	28.0	male	33.000	3	no	southeast	4449.4620
3	4	33.0	male	22.705	0	no	northwest	21984.4700
4	5	32.0	male	28.880	0	no	northwest	3866.8552
•••								
1333	1334	50.0	male	30.970	3	no	northwest	10600.5480
1334	1335	18.0	female	31.920	0	no	northeast	2205.9807
1335	1336	18.0	female	36.850	0	no	southeast	1629.8335
1336	1337	21.0	female	25.800	0	no	southwest	2007.9450
1337	1338	61.0	female	29.070	0	yes	northwest	29141.3600

1338 rows × 8 columns

#!pip install pandas-profiling

#pip install -U ydata-profiling

→ EDA

```
# from pandas_profiling import ProfileReport
'''from ydata_profiling import ProfileReport
ProfileReport(df)'''
      'from ydata_profiling import ProfileReport\nProfileReport(df)'
df.dropna(subset=["age","bmi","smoker","region"],inplace = True)
df.drop_duplicates(inplace=True)
print(df)
                                               children smoker
            person_id
                                 sex
                                          bmi
                                                                     region \
                        age
     0
                    1
                       19.0
                              female 27.900
                                                      0
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                                                                 southwest
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                                male 33.770
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     2
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                       18.0
                              female 31.920
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                 1336
                       18.0
                              female 36.850
                                                      0
                                                                 southeast
                                                             no
     1336
                 1337
                       21.0
                              female 25.800
                                                      0
                                                                 southwest
                                                             no
                              female 29.070
                                                      0
     1337
                 1338 61.0
                                                            yes
                                                                northwest
            total_charges
     0
               16884.9240
     1
                1725.5522
     2
                4449.4620
     3
               21984.4700
     4
                3866.8552
               10600.5480
     1333
     1334
                2205.9807
     1335
                1629.8335
     1336
                2007.9450
     1337
               29141.3600
     [1328 rows x 8 columns]
    Encoding
df['sex'].unique()
     array(['female', 'male'], dtype=object)
df['sex'] = df['sex'].replace('male',0)
df['sex'] = df['sex'].replace('female',1)
```

df['smoker'].unique()

array(['yes', 'no'], dtype=object)

```
df['smoker'] = df['smoker'].replace('no',0)
df['smoker'] = df['smoker'].replace('yes',1)
df['region'].unique()
     array(['southwest', 'southeast', 'northwest', 'northeast'], dtype=object)
df['region'] = df['region'].replace('southwest',0)
df['region'] = df['region'].replace('southeast',1)
df['region'] = df['region'].replace('northwest',2)
df['region'] = df['region'].replace('northeast',3)
Train Test data
# Selecting only 'age' and 'smoker' columns from the DataFrame
x=df.drop(columns='total_charges')
y=df.total_charges
# Split the data into train and test
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, random_state=0)
#Standardise data
from sklearn.preprocessing import StandardScaler
st=StandardScaler()
st.fit(x_train)
st.fit(x_test)
x_train_std=st.fit_transform(x_train)
x test std=st.fit transform(x test)
x_train_std.shape
     (996, 7)
!pip install livelossplot
     Collecting livelossplot
       Downloading livelossplot-0.5.5-py3-none-any.whl (22 kB)
     Requirement already satisfied: matplotlib in /usr/local/lib/python3.10/dist-r
```

Downloading livelossplot-0.5.5-py3-none-any.whl (22 kB)
Requirement already satisfied: matplotlib in /usr/local/lib/python3.10/dist-packag
Requirement already satisfied: bokeh in /usr/local/lib/python3.10/dist-packag
Requirement already satisfied: Jinja2>=2.9 in /usr/local/lib/python3.10/distRequirement already satisfied: contourpy>=1 in /usr/local/lib/python3.10/distRequirement already satisfied: numpy>=1.16 in /usr/local/lib/python3.10/distRequirement already satisfied: packaging>=16.8 in /usr/local/lib/python3.10/distRequirement already satisfied: pandas>=1.2 in /usr/local/lib/python3.10/distRequirement already satisfied: pillow>=7.1.0 in /usr/local/lib/python3.10/distRequirement already satisfied: PyYAML>=3.10 in /usr/local/lib/python3.10/distRequirement already satisfied: xyzservices>=2021.09.1 in /usr/local/lib/python3.10/distRequirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.10/
Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.10/
Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.10/

Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3 Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/c Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-pac Installing collected packages: livelossplot Successfully installed livelossplot-0.5.5

```
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Input, Dense
from livelossplot import PlotLossesKerasTF
```

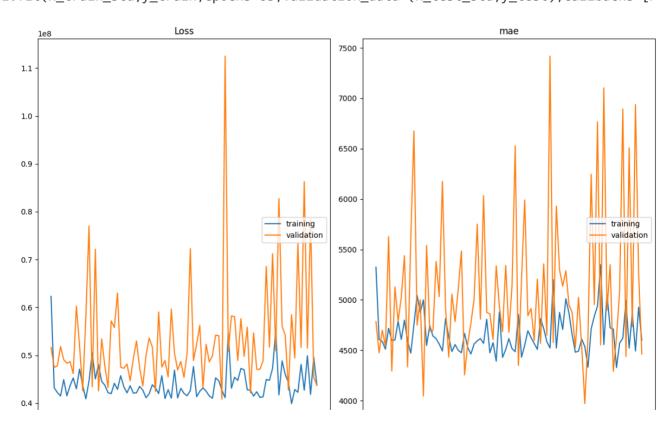
Design Perceptron

```
# Perceptron
model=Sequential()

# Add input and hidden Layer
model.add(Input(shape=(7,),name="Input Layer"))
model.add(Dense(1,activation='relu',name='OutputLayer'))

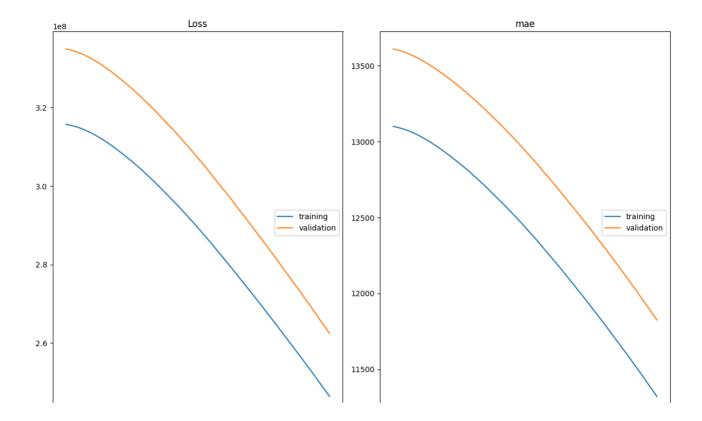
# compile the model
opt = tf.keras.optimizers.SGD(learning_rate=0.1,momentum=0.9,nesterov=True)  # momentum=
model.compile(loss='mse',optimizer=opt,metrics=['mae'])

# train the model
model.fit(x_train_std,y_train,epochs=85,validation_data=(x_test_std,y_test),callbacks=[!]
```



Design Deep Neural Network

```
'''model1 = Sequential()
model1.add(Input(shape=(7,),name="Input Layer"))
model1.add(Dense(14, activation='relu', name="Hidden1"))
model1.add(Dense(21, activation='relu', name="Hidden2"))
model1.add(Dense(1, activation='relu', name='OutputLayer'))
# Compile the model
opt = tf.keras.optimizers.SGD(learning_rate=0.01, momentum=0.9)
model1.compile(loss='mse', optimizer=opt, metrics=['mae'])
# Train the model
model1.fit(x_train_std,y_train,epochs=85,validation_data=(x_test_std,y_test),callbacks=
model1 = Sequential()
model1.add(Input(shape=(7,),name="Input Layer"))
model1.add(Dense(14, activation='relu', name="Hidden1"))
model1.add(Dense(21, activation='relu', name="Hidden2"))
model1.add(Dense(1, activation='relu', name='OutputLayer'))
# Compile the model
opt = tf.keras.optimizers.Adagrad(learning_rate=0.01)
model1.compile(loss='mse', optimizer=opt, metrics=['mae'])
# Train the model
```



```
1296
        1708.9258
        21797.0000
455
468
        23288.9280
261
        17085.2680
         4544.2350
1260
         9377.9040
1327
1285
         8534.6720
589
         5976.8310
1025
         2020.1770
         3490.5490
141
Name: total_charges, Length: 332, dtype: float64
```

```
# save model
model1.save("insurance.h5")
```

/usr/local/lib/python3.10/dist-packages/keras/src/engine/training.py:3103: Us
saving_api.save_model(