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# Machine Learning

**Lab Da-3**

**Single Layer Perceptron**

**and KNN Classifier**

**Dataset:**

| | Year | Model | Price | buy |
|---|---|---|---|---|
| 0 | 2010 | 1.10 | 120 | no |
| 1 | 2011 | 1.20 | 150 | no |
| 2 | 2011 | 1.80 | 160 | no |
| 3 | 2012 | 2.00 | 145 | no |
| 4 | 2012 | 2.10 | 150 | no |
| 5 | 2012 | 2.40 | 160 | no |
| 6 | 2013 | 2.60 | 190 | no |
| 7 | 2013 | 2.80 | 210 | no |
| 8 | 2014 | 2.90 | 220 | no |
| 9 | 2014 | 3.00 | 200 | no |
| 10 | 2015 | 3.10 | 210 | no |
| 11 | 2015 | 3.20 | 215 | yes |
| 12 | 2016 | 3.30 | 220 | yes |
| 13 | 2016 | 3.40 | 260 | no |
| 14 | 2017 | 3.50 | 270 | no |
| 15 | 2017 | 3.70 | 280 | no |
| 16 | 2018 | 4.00 | 250 | yes |
| 17 | 2018 | 4.20 | 264 | yes |
| 18 | 2018 | 4.40 | 270 | yes |
| 19 | 2019 | 4.60 | 282 | no |
| 20 | 2019 | 4.70 | 295 | no |
| 21 | 2020 | 4.80 | 303 | no |
| 22 | 2020 | 4.82 | 312 | no |
| 23 | 2020 | 4.90 | 323 | yes |
| 24 | 2021 | 4.95 | 337 | yes |
| 25 | 2021 | 5.00 | 300 | yes |
| 26 | 2022 | 5.10 | 315 | yes |
| 27 | 2022 | 5.20 | 340 | yes |
| 28 | 2022 | 5.40 | 360 | no |
| 29 | 2023 | 5.80 | 399 | yes |

## Q1) Apply KNN classification on the dataset and visualize the same

## Making a function to check the distances of an input point from all points in the dataset

```
def knn(year,price):
    df['diff1']=df['Year']-year
    df['diff2']=df['Price']-price
    df['d1sqr']=df['diff1']**2
    df['d2sqr']=df['diff2']**2
    df['dist']=np.sqrt(df['d2sqr']+df['d1sqr'])
    print(df)
```

## eg:

```
knn(2017,340)
```

| | Year | Model | Price | buy | diff1 | diff2 | d1sqr | d2sqr | dist |
|---|---|---|---|---|---|---|---|---|---|
| 0 | 2010 | 1.10 | 120 | no | -7 | -220 | 49 | 48400 | 220.111335 |
| 1 | 2011 | 1.20 | 150 | no | -6 | -190 | 36 | 36100 | 190.094713 |
| 2 | 2011 | 1.80 | 160 | no | -6 | -180 | 36 | 32400 | 180.099972 |
| 3 | 2012 | 2.00 | 145 | no | -5 | -195 | 25 | 38025 | 195.064092 |
| 4 | 2012 | 2.10 | 150 | no | -5 | -190 | 25 | 36100 | 190.065778 |
| 5 | 2012 | 2.40 | 160 | no | -5 | -180 | 25 | 32400 | 180.069431 |
| 6 | 2013 | 2.60 | 190 | no | -4 | -150 | 16 | 22500 | 150.053324 |
| 7 | 2013 | 2.80 | 210 | no | -4 | -130 | 16 | 16900 | 130.061524 |
| 8 | 2014 | 2.90 | 220 | no | -3 | -120 | 9 | 14400 | 120.037494 |
| 9 | 2014 | 3.00 | 200 | no | -3 | -140 | 9 | 19600 | 140.032139 |
| 10 | 2015 | 3.10 | 210 | no | -2 | -130 | 4 | 16900 | 130.015384 |
| 11 | 2015 | 3.20 | 215 | yes | -2 | -125 | 4 | 15625 | 125.015999 |
| 12 | 2016 | 3.30 | 220 | yes | -1 | -120 | 1 | 14400 | 120.004167 |
| 13 | 2016 | 3.40 | 260 | no | -1 | -80 | 1 | 6400 | 80.006250 |
| 14 | 2017 | 3.50 | 270 | no | 0 | -70 | 0 | 4900 | 70.000000 |
| 15 | 2017 | 3.70 | 280 | no | 0 | -60 | 0 | 3600 | 60.000000 |
| 16 | 2018 | 4.00 | 250 | yes | 1 | -90 | 1 | 8100 | 90.005555 |
| 17 | 2018 | 4.20 | 264 | yes | 1 | -76 | 1 | 5776 | 76.006579 |
| 18 | 2018 | 4.40 | 270 | yes | 1 | -70 | 1 | 4900 | 70.007142 |
| 19 | 2019 | 4.60 | 282 | no | 2 | -58 | 4 | 3364 | 58.034473 |
| 20 | 2019 | 4.70 | 295 | no | 2 | -45 | 4 | 2025 | 45.044423 |
| 21 | 2020 | 4.80 | 303 | no | 3 | -37 | 9 | 1369 | 37.121422 |
| 22 | 2020 | 4.82 | 312 | no | 3 | -28 | 9 | 784 | 28.160256 |
| 23 | 2020 | 4.90 | 323 | yes | 3 | -17 | 9 | 289 | 17.262677 |
| 24 | 2021 | 4.95 | 337 | yes | 4 | -3 | 16 | 9 | 5.000000 |
| 25 | 2021 | 5.00 | 300 | yes | 4 | -40 | 16 | 1600 | 40.199502 |
| 26 | 2022 | 5.10 | 315 | yes | 5 | -25 | 25 | 625 | 25.495098 |
| 27 | 2022 | 5.20 | 340 | yes | 5 | 0 | 25 | 0 | 5.000000 |
| 28 | 2022 | 5.40 | 360 | no | 5 | 20 | 25 | 400 | 20.615528 |
| 29 | 2023 | 5.80 | 399 | yes | 6 | 59 | 36 | 3481 | 59.304300 |

## Make function that take input values and k value and gives the classification of that point according to k nearest neighbours

```
def knn_classify(year,price,k):
    df['diff1']=df['Year']-year
    df['diff2']=df['Price']-price
    df['d1sqr']=df['diff1']**2
    df['d2sqr']=df['diff2']**2
    df['distance']=np.sqrt(df['d2sqr']+df['d1sqr'])
    top_k = df.nsmallest(k, 'distance')
    prediction = top_k['buy'].value_counts().head(1).index[0]
    return prediction
```

## returns yes/no

```
knn_classify(2017,340,5)
```

```
knn_classify(2017,340,5)
✓ 0.1s
'yes'

knn_classify(2023,420,5)
✓ 0.1s
'yes'

knn_classify(2016,200,5)
✓ 0.1s
'no'

knn_classify(2016,200,7)
✓ 0.2s
'no'
```

## Plotting the dataset and plotting regions according to what classification they would give

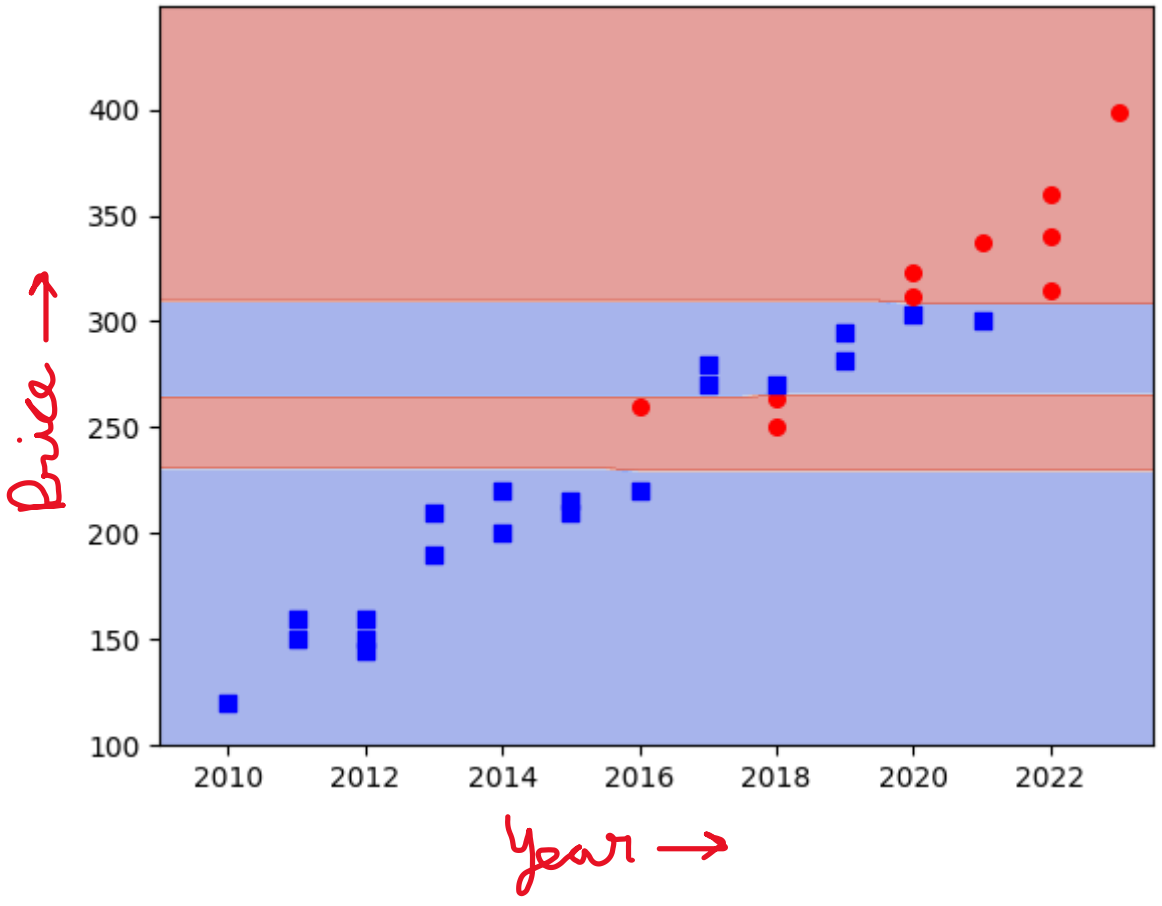
```
k=5
xx, yy = np.meshgrid(np.arange(2009, 2024, 0.5), np.arange(100, 450, 1))
zz = np.array([knn_classify(x, y,k)=='yes' for x, y in zip(xx.ravel(), yy.ravel())])
zz = zz.reshape(xx.shape)
# print(zz)
fig, ax = plt.subplots()
ax.contourf(xx, yy, zz, cmap='coolwarm', alpha=0.5)

colors = {'yes': 'red', 'no': 'blue'}
markers = {'yes': 'o', 'no': 's'}
for i, row in df.iterrows():
    ax.scatter(row['Year'], row['Price'], c=colors[knn_classify(row['Year'],
row['Price'],k)], marker=markers[knn_classify(row['Year'], row['Price'],k)], s=30)
plt.show()
```

plot the contour

Make contours according to classification and color them accordingly

**Red="Yes" Blue="No"**



## Q2) Pass the dataset through a single layer perceptron to find the weights of each feature and the bias

## Converting class labels “yes” and “no” to ordinal 0 and 1

```
df['ifBuy']=(df['buy'].map({'yes':1,'no':0}))
```

| | Year | Model | Price | buy | ifBuy |
|---|---|---|---|---|---|
| 0 | 2010 | 1.10 | 120 | no | 0 |
| 1 | 2011 | 1.20 | 150 | no | 0 |
| 2 | 2011 | 1.80 | 160 | no | 0 |
| 3 | 2012 | 2.00 | 145 | no | 0 |
| 4 | 2012 | 2.10 | 150 | no | 0 |
| 5 | 2012 | 2.40 | 160 | no | 0 |
| 6 | 2013 | 2.60 | 190 | no | 0 |
| 7 | 2013 | 2.80 | 210 | no | 0 |
| 8 | 2014 | 2.90 | 220 | no | 0 |
| 9 | 2014 | 3.00 | 200 | no | 0 |
| 10 | 2015 | 3.10 | 210 | no | 0 |
| 11 | 2015 | 3.20 | 215 | yes | 1 |
| 12 | 2016 | 3.30 | 220 | yes | 1 |
| 13 | 2016 | 3.40 | 260 | no | 0 |
| 14 | 2017 | 3.50 | 270 | no | 0 |
| 15 | 2017 | 3.70 | 280 | no | 0 |
| 16 | 2018 | 4.00 | 250 | yes | 1 |
| 17 | 2018 | 4.20 | 264 | yes | 1 |
| 18 | 2018 | 4.40 | 270 | yes | 1 |
| 19 | 2019 | 4.60 | 282 | no | 0 |
| 20 | 2019 | 4.70 | 295 | no | 0 |
| 21 | 2020 | 4.80 | 303 | no | 0 |
| 22 | 2020 | 4.82 | 312 | no | 0 |
| 23 | 2020 | 4.90 | 323 | yes | 1 |
| 24 | 2021 | 4.95 | 337 | yes | 1 |
| 25 | 2021 | 5.00 | 300 | yes | 1 |
| 26 | 2022 | 5.10 | 315 | yes | 1 |
| 27 | 2022 | 5.20 | 340 | yes | 1 |
| 28 | 2022 | 5.40 | 360 | no | 0 |
| 29 | 2023 | 5.80 | 399 | yes | 1 |

**Running a single layer perceptron with all weights starting at 1, alpha as 0.005 an for 10000 epochs**

```
w0=1
w1=1
w2=1
w3=1
alpha=0.005
epochs=10000

for i in range(0,epochs):
    for j in range(0,df.shape[0]):
        sum=w0*(-100)+w1*df['Year'][j]+w2*df['Model'][j]+w3*df['Price'][j]

        if sum>=0:
            pred=1
        else:
            pred=0
        delta=alpha*(df['ifBuy'][j]-pred)

        w0=w0-delta
        w1=w1+delta*df['Year'][j]
        w2=w2+delta*df['Model'][j]
        w3=w3+delta*df['Price'][j]

df['preds']=w0+w1*df['Year']+w2*df['Model']+w3*df['Price']
df['preds']=(df['preds'] >=0).astype(int)
print(df)
print(w0,w1,w2,w3)
```

initial weights

Activation function

update weights

| | Year | Model | Price | buy | ifBuy | preds |
|---|---|---|---|---|---|---|
| 0 | 2010 | 1.10 | 120 | no | 0 | 0 |
| 1 | 2011 | 1.20 | 150 | no | 0 | 0 |
| 2 | 2011 | 1.80 | 160 | no | 0 | 0 |
| 3 | 2012 | 2.00 | 145 | no | 0 | 0 |
| 4 | 2012 | 2.10 | 150 | no | 0 | 0 |
| 5 | 2012 | 2.40 | 160 | no | 0 | 0 |
| 6 | 2013 | 2.60 | 190 | no | 0 | 0 |
| 7 | 2013 | 2.80 | 210 | no | 0 | 0 |
| 8 | 2014 | 2.90 | 220 | no | 0 | 0 |
| 9 | 2014 | 3.00 | 200 | no | 0 | 0 |
| 10 | 2015 | 3.10 | 210 | no | 0 | 0 |
| 11 | 2015 | 3.20 | 215 | yes | 1 | 0 |
| 12 | 2016 | 3.30 | 220 | yes | 1 | 0 |
| 13 | 2016 | 3.40 | 260 | no | 0 | 0 |
| 14 | 2017 | 3.50 | 270 | no | 0 | 0 |
| 15 | 2017 | 3.70 | 280 | no | 0 | 0 |
| 16 | 2018 | 4.00 | 250 | yes | 1 | 0 |
| 17 | 2018 | 4.20 | 264 | yes | 1 | 0 |
| 18 | 2018 | 4.40 | 270 | yes | 1 | 0 |
| 19 | 2019 | 4.60 | 282 | no | 0 | 0 |
| 20 | 2019 | 4.70 | 295 | no | 0 | 0 |
| 21 | 2020 | 4.80 | 303 | no | 0 | 0 |
| 22 | 2020 | 4.82 | 312 | no | 0 | 0 |
| 23 | 2020 | 4.90 | 323 | yes | 1 | 0 |
| 24 | 2021 | 4.95 | 337 | yes | 1 | 0 |
| 25 | 2021 | 5.00 | 300 | yes | 1 | 0 |
| 26 | 2022 | 5.10 | 315 | yes | 1 | 0 |
| 27 | 2022 | 5.20 | 340 | yes | 1 | 0 |
| 28 | 2022 | 5.40 | 360 | no | 0 | 0 |
| 29 | 2023 | 5.80 | 399 | yes | 1 | 1 |

1.0349999999999993 -2.3549999999977373 32.28059999998488 11.830000000010772

$W_0$ $W_1$ $W_2$ $W_3$

**Clearly, a single layer perceptron provides a very bad fit for the given dataset even after 10000 epochs, as it has 3 features and 30 samples.**

**More layers would be required to get any substantial fit in a neural network.**