

You have just triggered a mouse gesture. Would you like to learn more about mouse gestures?

## jupyter Untitled8 Last Checkpoint: 9 minutes ago (unsaved changes)

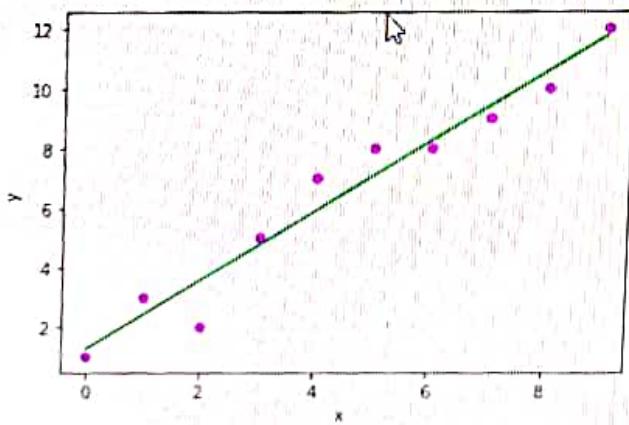
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Cell Kernel Help

```
# plotting regression line
plot_regression_line(x, y, b)

if __name__ == "__main__":
    main()

Estimated coefficients:
b_0 = 1.2363636363636363
b_1 = 1.1696969696969697
```



In [ ]:

Type here to search



## Experiment - 6

Implement linear regression using python

```
import numpy as np
import matplotlib.pyplot as plt
def estimate_coef(x, y):
    # number of observations / points
    n = np.size(x)
    # mean of x and y vector
    m_x = np.mean(x)
    m_y = np.mean(y)
    # Calculating cross-deviation and deviation about x
    SS_xy = np.sum(y * x) - n * m_y * m_x
    SS_xx = np.sum(x * x) - n * m_x * m_x
    # calculating regression coefficients
    b_1 = SS_xy / SS_xx
    b_0 = m_y - b_1 * m_x
    return (b_0, b_1)

def plot_regression_line(x, y, b):
    # plotting the actual points as scatter plot
    plt.scatter(x, y, color="m", marker="o", s=30)
    # predicted response vector
    y_pred = b[0] + b[1] * x
```

```
#F plotting the regression line
plt.plot(x,y-pred,color="g")
```

```
#F putting labels
```

```
plt.xlabel('x')
```

```
plt.ylabel('y')
```

```
# function to show plot
```

```
plt.show()
```

```
def main():
```

```
# observations / data
```

```
x=np.array([0,1,2,3,4,5,6,7,8,9])
```

```
y=np.array([1,3,2,5,7,8,8,9,10,12])
```

```
# estimating coefficients
```

```
b=estimate_coef(x,y)
```

```
Point ("Estimated coefficients : In b[0] = {} \\\n      nb[1] = {} ".format(b[0],b[1]))
```

```
# plotting regression line
```

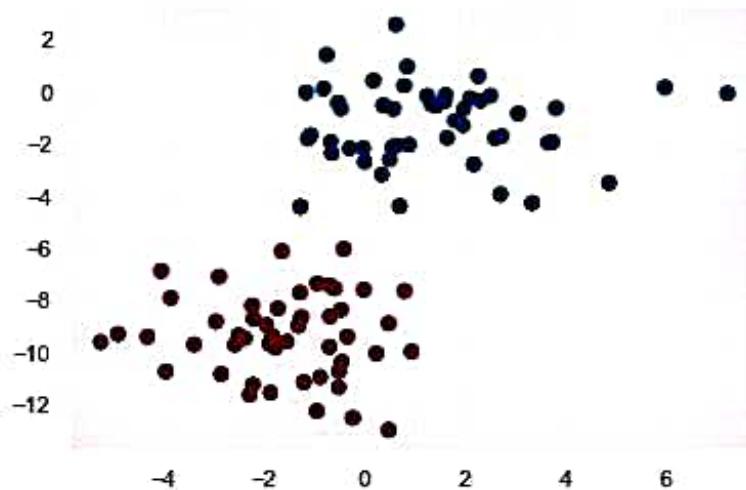
```
plot_regression_line(x,y,b)
```

```
if __name__ == "__main__":
```

```
main()
```

```
In [6]: %matplotlib inline
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns; sns.set()
from sklearn.datasets import make_blobs
X, y = make_blobs(100, 2, centers=2, random_state=2, cluster_std=1.5)
plt.scatter(X[:, 0], X[:, 1], c=y, s=50, cmap='RdBu')
```

```
Out[6]: <matplotlib.collections.PathCollection at 0x222116037f0>
```



## Experiment 7

Example 1:

Implement Naive Bayes theorem to classify the english text.

Source Code:

```
%matplotlib inline  
import numpy as np  
import matplotlib.pyplot as plt  
import seaborn as sns; sns.set()  
from sklearn.datasets import make_blobs  
x,y = make_blobs(100,2,centers=2,random_state=2,  
                  cluster_std=1.5)  
plt.scatter(x[:,0], x[:,1], c=y, s=50, cmap='RdBu')
```

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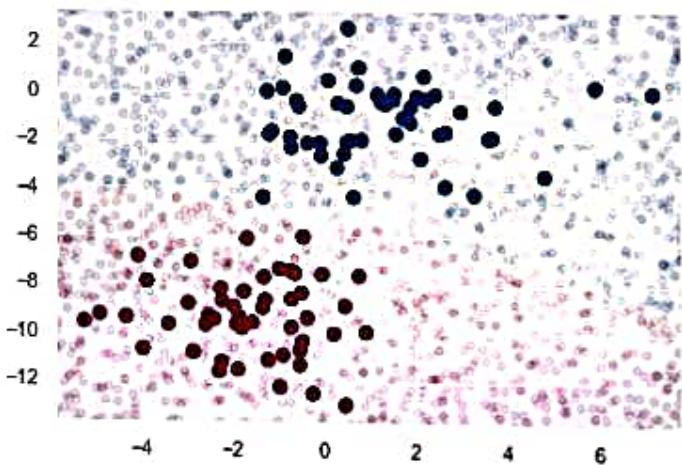
## jupyter Untitled3 Last Checkpoint: 10 minutes ago (autosaved)

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Code

```
yprob = model.predict_proba(xnew)  
yprob[-8:].round(2)
```

```
Out[7]: array([[0.89, 0.11],  
               [1. , 0. ],  
               [1. , 0. ],  
               [1. , 0. ],  
               [1. , 0. ],  
               [1. , 0. ],  
               [0. , 1. ],  
               [0.15, 0.85]])
```



Example 2:

Implement Naïve Bayes theorem to classify the english text

Source code :

```
%matplotlib inline
```

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

```
import seaborn as sns; sns.set()
```

```
from sklearn.datasets import make_blobs
```

```
x,y = make_blobs(100, 2, centers=2, random_state=2, cluster_std=1.5)
```

```
plt.scatter(x[:,0], x[:,1], c=y, s=50, cmap='RdBu');
```

```
from sklearn.naive_bayes import GaussianNB
```

```
model = GaussianNB()
```

```
model.fit(x,y);
```

```
rng = np.random.RandomState(0)
```

```
Xnew = [-6, -14] + [14, 18]*rng.rand(2000, 2)
```

```
ynew = model.predict(Xnew)
```

```
plt.scatter(x[:,0], x[:,1], c=y, s=50, cmap='RdBu')
```

```
lim = plt.axis()
```

```
plt.scatter(Xnew[:,0], Xnew[:,1], c=ynew, s=20, cmap='RdBu', alpha=0.1)
```

```
plt.axis(lim);
```

```
yprob = model.predict_proba(Xnew)
```

```
yprob[-8: ].round(2)
```

localhost:8889/notebooks/Untitled3.ipynb

AliExpress

## Cell Untitled3 Last Checkpoint: 12 minutes ago (unsaved changes)

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```
"__nbits = 20
# define the population size
n_pop = 100
# crossover rate
r_cross = 0.9
# mutation rate
r_mut = 1.0 / float(n_bits)
# perform the genetic algorithm search
best, score = genetic_algorithm(onemax, n_bits, n_iter, n_pop, r_cross, r_mut)
print('Done!')
print('f(%s) = %f' % (best, score))

>0, new best f([1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1]) = -15.000
>1, new best f([1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0]) = -16.000
>3, new best f([1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1]) = -17.000
>4, new best f([1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1]) = -18.000
>4, new best f([1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1]) = -19.000
>7, new best f([1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1]) = -20.000000
Done!
f([1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1]) = -20.000000
```

[ ]:

## Experiment 8:

//genetic algorithm to evaluate a binary string based on the number of 1's in the string. a bitstring with a length of 20 bits will have a score of 20 for a string of all 1's in the string.

$$\text{111111111111111111} = 20, \text{111111110000000000} = 10$$

```
from numpy.random import randint
```

```
from numpy.random import rand
```

//objective function

```
def onemax(x):
```

```
    return -sum(x)
```

# tournament selection

```
def selection(pop, scores, k=3):
```

//first random selection

```
selection_ix = randint(len(pop))
```

```
for ix in randint(0, len(pop), k-1):
```

//checks if better (e.g. perform a tournament)

```
if scores[ix] < scores[selection_ix]:
```

```
    selection_ix = ix
```

```
return pop[selection_ix]
```

//crossover two parents to create two children

```
def crossover(p1, p2, r_cross):
```

//children are copies of parents by default

```
C1, C2 = p1.copy(), p2.copy()
```

//check for recombination

```
if rand() < r_cross:
```

```

# select crossover point that is not on the end of the
string.
pt = randint(1, len(p1) - 2)

# perform crossover
c1 = p1[:pt] + p2[pt:]
c2 = p2[:pt] + p1[pt:]

return [c1, c2]

# mutation operator
def mutation(bitstring, r_mut):
    for i in range(len(bitstring)):
        # check for a mutation
        if random() < r_mut:
            # flip the bit
            bitstring[i] = 1 - bitstring[i]

# genetic algorithm
def genetic_algorithm(objective, n_bits, n_iter, n_pop,
                      r_cross, r_mut):

    # initial population of random bitstring
    pop = [randint(0, 2, n_bits).tolist() for _ in range(n_pop)]

    # keep track of best solution
    best, best_eval = 0,
    objective(pop[0])

    # enumerate generations
    for gen in range(n_iter):
        # evaluate all candidates in the population
        scores = [objective(c) for c in pop]

```

# check for new best solution

for i in range (n-pop):

if & scores[i] < best\_eval:

best, best\_eval = pop[i];

scores[i]

Print (">-l.d, new best f(·)·s) = -l.3f".l. (gen, pop[i], scores[i])

# select parents

selected = [selection (pop, scores) for \_ in range (n-pop)]

# create the next generation

children = list()

for i in range (0, n-pop, 2):

# get selected parents in pairs

$P_1, P_2 = \text{selected}[i], \text{selected}[i+1]$

# crossover and mutation

for c in crossover ( $P_1, P_2, r_{\text{cross}}$ ):

# mutation

mutation (c, r\_mut)

# store for next generation

children.append(c)

# replace population

pop = children

return [best, best\_eval]

# define the total iterations

n\_iter = 100

# bits

n\_bits = 20

#define the population size

n-pop = 100

# crossover rate

r-cross = 0.9

# mutation rate

r-mut = 1.0 / float (n-bits)

#perform the genetic algorithm search

best, score = genetic-algorithm (one max, n-bits, n-iter,  
n-pop, r-cross, r-mut)

Print ('Done !')

Print ('f(-1.5) = ' + str(best))

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azon.in AliExpress

## jupyter Untitled3 Last Checkpoint: 27 minutes ago (autosaved)

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Code

```
>epoch=0, lrate=0.500, error=6.350
>epoch=1, lrate=0.500, error=5.531
>epoch=2, lrate=0.500, error=5.221
>epoch=3, lrate=0.500, error=4.951
>epoch=4, lrate=0.500, error=4.519
>epoch=5, lrate=0.500, error=4.173
>epoch=6, lrate=0.500, error=3.835
>epoch=7, lrate=0.500, error=3.506
>epoch=8, lrate=0.500, error=3.192
>epoch=9, lrate=0.500, error=2.898
>epoch=10, lrate=0.500, error=2.626
>epoch=11, lrate=0.500, error=2.377
>epoch=12, lrate=0.500, error=2.153
>epoch=13, lrate=0.500, error=1.953
>epoch=14, lrate=0.500, error=1.774
>epoch=15, lrate=0.500, error=1.614
>epoch=16, lrate=0.500, error=1.472
>epoch=17, lrate=0.500, error=1.346
>epoch=18, lrate=0.500, error=1.233
>epoch=19, lrate=0.500, error=1.132
[{'weights': [-1.4688375095432327, 1.850887325439514, 1.0858178629550297], 'output': 0.029980305604426185, 'delta': -0.0059546604162323625}, {'weights': [0.37711098142462157, -0.0625909894552989, 0.2765123702642716], 'output': 0.9456229000211323, 'delta': 0.0026279652850863837}], [{"weights": [2.515394649397849, -0.3391927502445985, -0.9671565426390275], "output": 0.23648794202357587, "delta": -0.04270059278364587}, {"weights": [-2.5584149848484263, 1.0036422106209202, 0.42383086467582715], "output": 0.7790535202438367, "delta": 0.03803132596437354}]
```

## Experiment 9:

Implement the finite words classification system using Back Propagation algorithm.

Source code:

```
from math import exp
```

```
from random import seed
```

```
from random import random
```

```
import matplotlib.pyplot as plt
```

# initialize a network

```
def initialize_network(n_inputs, n_hidden, n_outputs):
```

```
    network = list()
```

```
    hidden_layer = [ {'weights': [random() for i in range(n_inputs+1)]} ]
```

```
    for i in range(n_hidden)
```

```
        network.append(hidden_layer)
```

```
    output_layer = [ {'weights': [random() for i in range(n_hidden+1)]} ]
```

```
    for i in range(n_outputs)
```

```
        network.append(output_layer)
```

```
    return network
```

# calculate neuron activation for an input

```
def activate(weights, inputs):
```

```
    activation = weights[-1]
```

```
    for i in range(len(weights)-1):
```

```
        activation += weights[i] * inputs[i]
```

```
    return activation
```

```

## Transfer neuron activation
def transfer(activation):
    return 1.0 / (1.0 + np.exp(-activation))

## Forward propagate input to a network output
def forward_propagate(network, row):
    inputs = row
    for layer in network:
        new_inputs = []
        for neuron in layer:
            activation = activate(neuron['weights'], inputs)
            neuron['output'] = transfer(activation)
            new_inputs.append(neuron['output'])
        inputs = new_inputs
    return inputs

## calculate the derivative of an neuron output
def transfer_derivative(output):
    return output * (1.0 - output)

## Back propagate error and store in neurons
def backward_propagate_error(network, expected):
    for i in reversed(range(len(network))):
        layer = network[i]
        errors = list()
        if i == len(network) - 1:
            for j in range(len(layer)):
                error = 0.0
        else:
            for j in range(len(layer)):
                total_error = 0.0
                for neuron in network[i + 1]:
                    total_error += neuron['errors'] * neuron['weights'][j]
                error = transfer_derivative(layer[j]['output']) * total_error
        errors.append(error)
    network[i] = errors

```

```

for neuron in network[i+1]:
    error += (neuron['weights'][j] * neuron['delta'])
errors.append(error)
else:
    for j in range(len(layer)):
        neuron = layer[j]
        neuron['delta'] = errors[j] *
            transfer_derivative(neuron['output'])
# Update network weights with error
def update_weights(network, row, l_rate):
    for i in range(len(network)):
        inputs = row[i-1]
        if i == 0:
            inputs = [neuron['output'] for neuron in network[i-1]]
        for neuron in network[i]:
            neuron['weights'][j] += l_rate * neuron['delta'] * inputs[j]
            neuron['weights'][-1] += l_rate * neuron['delta']
# Train a network (network, train, l_rate, n_epoch, n_outputs):
for epoch in range(n_epoch):
    sum_error = 0
    for row in train:
        outputs = forward_propagate(network, row)
        expected = [0 for i in range(n_outputs)]
        expected[row[-1]] = 1
        sum_error += sum([(expected[i] - outputs[i]) ** 2

```

```

for i in range(len(expected))):
    backward-propagate-error(network, expected)
    update-weights(network, row, l-rate)
print('epoch = %.d, lrate = %.3f, error = %.3f' % (epoch,
    l_rate, sum-error))

```

# Test training backprop algorithm seed(1)

```

dataset = [
    [2.7810836, 2.550537003, 0],
    [1.465489372, 2.362125076, 0],
    [3.396561688, 4.400293529, 0],
    [1.38807019, 1.850220317, 0],
    [3.06407232, 3.005305973, 0],
    [7.627531214, 2.759262235, 1],
    [5.332441248, 2.088626775, 1],
    [6.922596716, 1.77106367, 1],
    [8.675418651, -0.242068655, 1],
    [7.673756466, 3.508563011, 1]
]

```

n\_inputs = len(dataset[0]) - 1

n\_outputs = len(set([row[-1] for row in dataset]))

network = initialize\_network(n\_inputs, 2, n\_outputs)

train\_network(network, dataset, 0.5, 20, n\_outputs)

for layer in network:

print(layer)

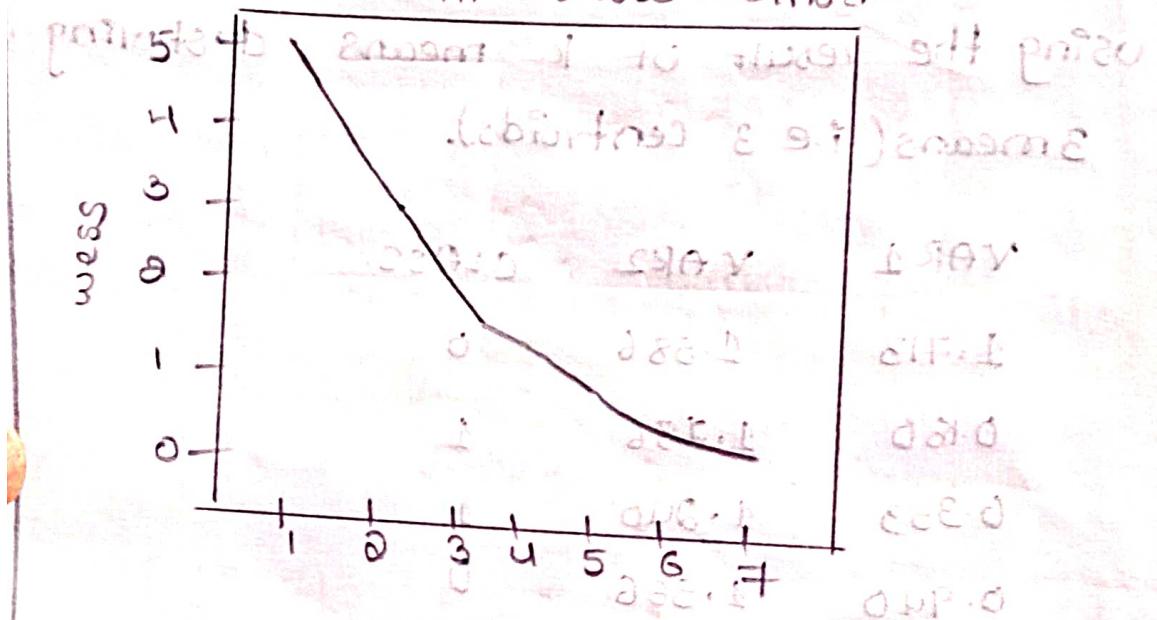
O/P [5.631208500060005, 2.487442083333333,

[1.754385325, 0.8482311666666667, 0.55849783

[4.338351560, 2.5424755000000005, 0.1953075000

not int[0,000008], a theory of k-means base

0.0 = Elbow method



O/P

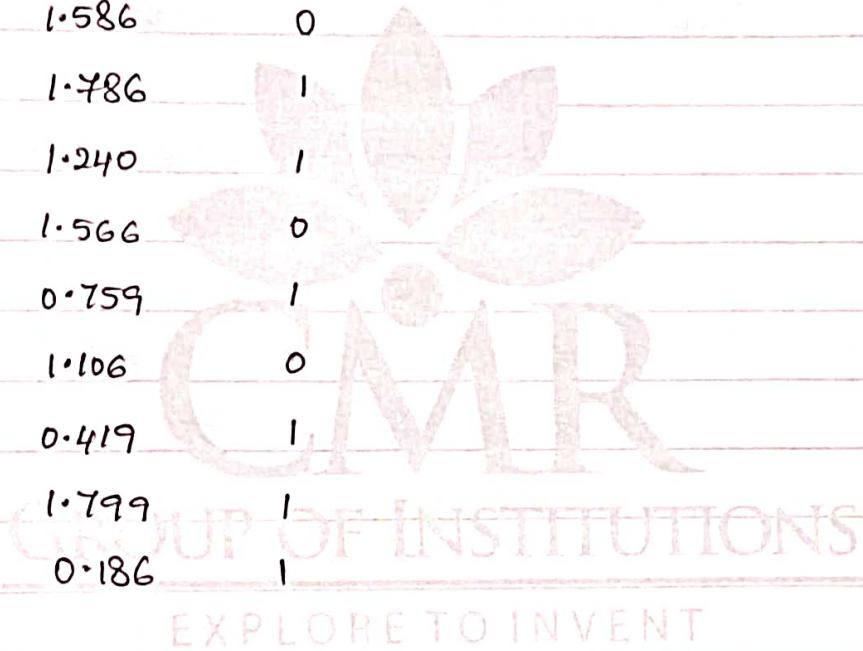
K-Means (n\_clusters=3, random\_state=0)

[0 1 1 1 2 0 1 2 1 2 2]

## Experiment - 1

Given the following data, which specify classifications for nine combinations of VAR1 and VAR2 predict a classification for a case where  $\text{VAR1} = 0.906$  and  $\text{VAR2} = 0.606$ , using the result of k-means clustering with 3 means (i.e., 3 centroids)

VAR1	VAR2	CLASS
1.713	1.586	0
0.180	1.786	1
0.353	1.240	1
0.940	1.566	0
1.486	0.759	1
1.266	1.106	0
1.540	0.419	1
0.459	1.799	1
0.773	0.186	1



Source code:

```
# importing the libraries
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd

# importing the Iris dataset with pandas
dataset = pd.read_csv('C:/Users/Desktop/IRIS-data.csv')
print(dataset)
```

`x = dataset.iloc [i, (1,2)] values  
print (x)`

#finding the optimum number of clusters for k means classification

`from sklearn.cluster import KMeans  
wess = []`

`for i in range (1,8):`

`Kmeans = Kmeans (n_clusters = i, init = k-means ++,  
max_iter = 300, n_init = 10, random_state = 0)`

`kmeans.fit(x)`

`wess.append (Kmeans.inertia_)`

`print (wess)`

# plotting the results onto a line graph, allowing us to observe 'The elbow'

`plt.plot (range (1,8), wess)`

`plt.title ('The elbow method')`

`plt.xlabel ('Number of clusters')`

`plt.ylabel ('Wess')`

`plt.show()`

output:

unconditional probability of golf is 0.4

conditional probability of single with medrisk:

status risk

married : highrisk 0.666667

single : lowrisk 0.666667

single : medrisk 0.333333

single : highrisk 0.666667

medrisk 0.666667

lowrisk 0.333333

(Unpivot Hofmort) after 5 iterations

purpose: to get better estimate

probability longfibrosis diff from previous

probability longfibrosis diff from Hof

Flagship software version 10.0

## Experiment - 5

The following training examples map descriptions of individuals onto high, medium and low credit-worthiness

medium	skiing	design	single	twenties	no → highrisk
high	golf	trading	married	forties	yes → lowrisk
low	Speedway	transport	married	thirties	yes → medrisk
medium	football	banking	single	thirties	yes → lowrisk
high	flying	media	married	fifties	yes → highrisk
low	football	security	single	twenties	no → medrisk
medium	golf	media	married	thirties	yes → medrisk
medium	golf	transport	single	forties	yes → lowrisk
high	skiing	banking	married	thirties	yes → highrisk
low	golf	unemployed	single	forties	yes → highrisk

Input attributes are (from left to right)  
 income, recreation, job, status, age-group, how-owner. find  
 the unconditional probability of 'golf' and the conditional  
 probability of 'single' given 'medRisk' in the dataset?

Source code:

```

import pandas as pd
import numpy as np
dataset = pd.read_csv('C:/Users/Desktop/credit.csv')
df = pd.DataFrame(dataset)
l = len(df)
  
```

```
rec = df[['recreation']].values.tolist()
```

c=0

for i in rec:

if (i == "golf"):

cl = 1

```
print(" unconditional probability of golf is:", cl);
```

cl = 0

```
risk = df[['risk']].values.tolist()
```

for i in the risk:

if (i == 'medrisk'):

clt = 1

```
Print(" Conditional probability of single with med risk:")
```

```
Print(df.groupby('status').visit['risk'].value_counts()  
() / cl)
```

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