

```
In [1]: import numpy as np
import pandas as pd

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import StratifiedKFold

import torch
import torch.nn as nn
import torch.optim as Adam
import matplotlib.pyplot as plt
from torch import nn
```

1. Baye's Theorem

M is a marker that determines genetic disposition to kidney disease. A chemical test can show if you are positive or negative for M. However, test is not 100% right. $y = +ve$ or $-ve$
 $x =$ marker (M) or no marker (no M)

$$P[+|M] = 0.95$$

$$P[-|no M] = 0.95$$

$$P[M] = 0.01$$

1a

$$P[-|M] = 0.05$$

$$P[+|not M] = 0.05$$

$$P[not M] = 0.99$$

1b

```
In [2]: have_M = 0.01
have_M_test_pos = 0.95

no_M = 0.95
no_M_test_pos = 0.05

actually_have_M = have_M * have_M_test_pos
false_positive = no_M * no_M_test_pos

all_positive_tests = actually_have_M + false_positive
odds_that_Korede_has_M = actually_have_M / all_positive_tests

print(f'The odds that Korede actually has M are {round(odds_that_Korede_has_M,
```

The odds that Korede actually has M are 0.1667.

Thanks to Baye's Theorem, I would not be too worried about actually having M. The feature of the data that accounts for this result is the occurrence of M in the population (0.01).

1c

New Scenario:

$$P[M] = 0.10$$

$$P[\text{not } M] = 0.90$$

```
In [3]: have_M = 0.1
have_M_test_pos = 0.95

no_M = 0.90
no_M_test_pos = 0.05

actually_have_M = have_M * have_M_test_pos
false_positive = no_M * no_M_test_pos

all_positive_tests = actually_have_M + false_positive
odds_that_Korede_has_M = actually_have_M / all_positive_tests

print(f'The odds that Korede actually has M are {round(odds_that_Korede_has_M,
```

The odds that Korede actually has M are 0.6786!

Omg my odds are so much higher now. I would be worried.

2. Gaussian Naive Bayes

$P(\text{cultivar} | X)$ - probability of cultivar given an attribute

$P(\text{wine attribute } x | \text{cultivar})$ - probability an attribute given a cultivar

```
In [22]: wines = pd.read_csv('../wines.csv')
# we don't need start assignment, so drop it
wines = wines.drop('Start assignment', axis = 1)
# rename ranking as Cultivar for clarity
wines = wines.rename(columns = {'ranking': 'Cultivar'})
# shuffle values in Cultivar
shuffled_cultivar = wines['Cultivar'].sample(frac = 1)
wines['Cultivar'] = shuffled_cultivar.values

#wines.head()
```

2a

```

In [5]: class NaiveBayesClassifier():
    def __init__(self):
        self.type_indices = {}    # store the indices of wines that belong to
        self.type_stats = {}      # store the mean and std of each cultivar
        self.ndata = 0
        self.trained = False

    @staticmethod #static methods are bound to a class and not to instances
    def gaussian(x,mean,std):
        # Gaussian probability density formula
        exponent = -((x - mean)/ 2 * std)**2
        return (1 / (std * np.sqrt(2 * np.pi))) * np.exp(exponent)

    @staticmethod
    def calculate_statistics(x_values):
        # Returns a list with length of input features. Each element is a tuple
        n_feats = x_values.shape[1]
        return [(np.average(x_values[:,n]),np.std(x_values[:,n])) for n in range(n_feats)]

    @staticmethod
    def calculate_prob(x_input,stats):
        """Calculate the probability that the input features belong to a specific class
        x_input: np.array shape(nfeatures)
        stats: list of tuple [(mean1,std1),(means2,std2),...]
        """
        init_prob = 1.0
        for i, (mean, std) in enumerate(stats):
            feature_P = NaiveBayesClassifier.gaussian(x_input[i], mean, std)
            init_prob *= feature_P
        return init_prob

    def fit(self,xs,ys):
        # Train the classifier by calculating the statistics of different features
        self.ndata = len(ys)
        for y in set(ys):
            type_filter = (ys==y)
            self.type_indices[y] = type_filter
            self.type_stats[y] = self.calculate_statistics(xs[type_filter])
        self.trained = True

    def predict(self,xs):
        # Do the prediction by outputting the class that has highest probability
        if len(xs.shape) > 1:
            print("Only accepts one sample at a time!")
        if self.trained:
            guess = None
            max_prob = 0
            #  $P(C|X) = P(X|C)*P(C) / \sum_i(P(X|C_i)*P(C_i))$  (denominator for all classes)
            for y_type in self.type_stats:
                prob = self.calculate_prob(xs, self.type_stats[y_type])
                if prob > max_prob:
                    max_prob = prob
                    guess = y_type
            # use to troubleshoot
            # print (f'max prob {max_prob}, variable prob {prob}')

```

```

        return guess
    else:
        print("Please train the classifier first!")

```

I chose this function form because all the attributes are continuous; it is appropriate to represent them using a gaussian function.

Now to find $P(\text{alcohol \% 13} \mid \text{Cultivar 1})$:

```

In [6]: # INITIATE CLASSIFIER
Classifier1 = NaiveBayesClassifier()

# SELECT FEATURES AND LABELS FROM DATAFRAME
x_values = wines.iloc[:, :13].values
y_values = wines['Cultivar'].values

Classifier1.fit(x_values, y_values)

# GIVEN MEAN AND STD OF CULTIVAR 1, FIND
# LIKELIHOOD OF IT HAVING ALCOHOL % OF 13
mean, std = Classifier1.type_stats[1][0]

P_of_13_given_cultivar_1 = Classifier1.gaussian(13, mean, std)
print(f'The probability of a wine from Cultivar 1 having an alcohol % of 13%

```

The probability of a wine from Cultivar 1 having an alcohol % of 13% is 0.5046355789936335

2b

```

In [7]: # NORMALIZE DATA
def normalize_data(df):
    return df.iloc[:, :-2].apply(lambda x: (x - x.mean()) / x.std())
    # "iloc[:, :-1]" to leave out Cultivar column

wines_normalized = normalize_data(wines.copy())
wines_normalized['Cultivar'] = wines['Cultivar']
#wines_normalized.head()

features = wines.drop(columns='Cultivar').values
scaler = StandardScaler()
x_norm = scaler.fit_transform(features)

```

In the next cell, divide the data into three groups (folds) and predict using Naive Bayes.

```

In [8]: n_folds = 3      # for 3-fold cross validation
train_sets = []
test_sets = []

# ITERATE THROUGH FOLDS
for i in range(n_folds):
    X_train_df, X_test_df, y_train_series, y_test = train_test_split(wines_r
    # variables on LHS are dataframes and series. Convert to numpy arrays
    X_train = X_train_df.values

```

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X_test = X_test_df.values
y_train = y_train_series.values
# add variables to their respective sets
train_sets.append((X_train, y_train))
test_sets.append((X_test, y_test))

# initiate model. fit and predict
NB_classifier = NaiveBayesClassifier()
NB_classifier.fit(X_train, y_train)
y_predictions = np.array([NB_classifier.predict(X_train[0])])

```

```

/var/folders/m8/skfw9g2x4_g4pq5cv80_g24w0000gn/T/ipykernel_8594/2531548476.p
y:12: RuntimeWarning: divide by zero encountered in scalar divide
return (1 / (std * np.sqrt(2 * np.pi))) * np.exp(exponent)

```

```

In [9]: # calculating accuracy of naive bayes
def calculate_accuracy(model,xs,ys):
    y_pred = np.zeros_like(ys)
    for idx,x in enumerate(xs):
        y_pred[idx] = model.predict(x)
    return np.sum(ys==y_pred)/len(ys)

print(f'The accuracy of the Naive Bayes classifier is {calculate_accuracy(NB_classifier,X_test,y_test)}')

```

The accuracy of the Naive Bayes classifier is 0.36666666666666664

```

/var/folders/m8/skfw9g2x4_g4pq5cv80_g24w0000gn/T/ipykernel_8594/2531548476.p
y:12: RuntimeWarning: divide by zero encountered in scalar divide
return (1 / (std * np.sqrt(2 * np.pi))) * np.exp(exponent)

```

I heard in lecture that Naive Bayes should be ~0.9 accuracy, so I know that it is more accurate than other methods but unfortunately was not able to demonstrate that in my code.

Naive Bayes outperforms Simulates Annealing (which was used in assignment 2, questions 2d and 2e), Naive Bayes performs much better; Simulated Annealing had an average accuracy of 0.56.

3. Softmax and Cross Entropy Loss

```

In [10]: # Define a simple neural network with softmax activation
class SimpleNN(torch.nn.Module):
    def __init__(self, input_size, num_classes):
        super(SimpleNN, self).__init__()
        self.fc = torch.nn.Linear(input_size, num_classes)

    def forward(self, x):
        x = self.fc(x)
        return torch.softmax(x, dim=0) # Apply softmax activation

# Same as above WITHOUT softmax activation
class SimpleNNWithoutSoftmax(torch.nn.Module):
    def __init__(self, input_size, num_classes):
        super(SimpleNNWithoutSoftmax, self).__init__()

```

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        self.fc = torch.nn.Linear(input_size, num_classes)

    def forward(self, x):
        x = self.fc(x)
        return x # No softmax activation

```

```

In [23]: # change featues and labels to tensors
wines_train_X = torch.tensor(x_values, dtype=torch.float32)
wines_train_y = torch.tensor(y_values, dtype=torch.long)
#x_values.shape

```

3a

The softmax activation function transforms the raw model outputs into a probability distribution (all observations add up to 1). Outputs are slightly different with each run, so I have printed each one 3 times.

```

In [31]: # Define inputs and model
input_size = 13 # number of features
num_classes = 3 # number of cultivars

model_w_softmax = SimpleNN(input_size, num_classes)
# change wine features to tensor!
#train_X = torch.tensor(x_values, dtype=torch.float32)
# Pass the wine features through the network once without backpropagation
with torch.no_grad():
    output = model_w_softmax(wines_train_X)
print(f"Output with softmax activation: {output}")

# Define the model without softmax
model_wo_softmax = SimpleNNWithoutSoftmax(input_size, num_classes)
# Pass the data through the network one without backpropagation and without
with torch.no_grad():
    output_no_softmax = model_wo_softmax(wines_train_X)

```

Output with softmax activation: tensor([[0.0000e+00, 1.0807e-12, 4.4635e-05],

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[0.0000e+00, 5.5316e-28, 5.1829e-16]])
```

```
In [13]: print(f"Output with softmax activation: {output}")
```

Output with softmax activation: tensor([[4.2039e-44, 4.7292e-33, 0.0000e+00],

[1.6539e-25, 0.0000e+00, 4.3836e-28],
 [2.8026e-44, 1.3507e-30, 5.7341e-42],
 [0.0000e+00, 8.9663e-19, 0.0000e+00],
 [0.0000e+00, 3.3238e-16, 0.0000e+00],
 [0.0000e+00, 1.9144e-21, 0.0000e+00],
 [8.4555e-28, 0.0000e+00, 4.0589e-31],
 [9.2486e-43, 1.0192e-31, 3.2765e-40],
 [7.8513e-33, 5.8742e-40, 1.8481e-31],
 [0.0000e+00, 1.5584e-18, 0.0000e+00],
 [7.4269e-44, 5.6274e-31, 2.4076e-41],
 [0.0000e+00, 4.0563e-25, 0.0000e+00],
 [2.6769e-34, 5.0802e-40, 9.6662e-35],
 [0.0000e+00, 1.2979e-27, 2.8026e-45],
 [0.0000e+00, 1.4065e-28, 5.6052e-45],
 [1.0106e-34, 1.7258e-39, 1.2031e-34],
 [7.0065e-45, 2.6186e-30, 1.7656e-43],
 [0.0000e+00, 2.3130e-19, 0.0000e+00],
 [0.0000e+00, 3.4431e-25, 0.0000e+00],
 [0.0000e+00, 5.1969e-20, 0.0000e+00],
 [2.4178e-23, 0.0000e+00, 1.7405e-23],
 [6.4583e-09, 0.0000e+00, 2.0893e-08],
 [3.8648e-14, 0.0000e+00, 1.2404e-11],
 [5.3257e-34, 1.5050e-39, 1.6959e-33],
 [1.5733e-12, 0.0000e+00, 4.0622e-15],
 [2.6604e-21, 0.0000e+00, 1.7749e-18],
 [2.5110e-14, 0.0000e+00, 3.8103e-14],
 [4.4282e-10, 0.0000e+00, 4.2538e-11],
 [4.7646e-24, 0.0000e+00, 1.1337e-21],
 [2.1216e-21, 0.0000e+00, 2.3315e-17],
 [6.9576e-35, 2.3262e-43, 7.8052e-43],
 [1.7150e-18, 0.0000e+00, 6.6716e-24],
 [4.8560e-08, 0.0000e+00, 1.3342e-07],
 [3.2532e-25, 0.0000e+00, 1.8203e-24],
 [1.0918e-08, 0.0000e+00, 5.3437e-08],
 [3.7898e-16, 0.0000e+00, 1.1330e-17],
 [1.1323e-09, 0.0000e+00, 2.9782e-09],
 [5.0063e-07, 0.0000e+00, 2.5501e-06],
 [2.8031e-08, 0.0000e+00, 1.9392e-07],
 [2.5581e-20, 0.0000e+00, 6.0158e-20],
 [1.0522e-06, 0.0000e+00, 3.2719e-06],
 [3.5392e-04, 0.0000e+00, 2.1233e-04],
 [1.5598e-18, 0.0000e+00, 4.2619e-16],
 [1.4866e-16, 0.0000e+00, 4.2824e-17],
 [1.5986e-18, 0.0000e+00, 5.7014e-20],
 [2.0585e-18, 0.0000e+00, 3.9054e-17],
 [3.1126e-18, 0.0000e+00, 4.5606e-19],
 [2.3932e-14, 0.0000e+00, 5.8462e-15],
 [2.8570e-09, 0.0000e+00, 1.0786e-07],
 [1.2341e-15, 0.0000e+00, 3.0575e-18],
 [6.1849e-08, 0.0000e+00, 5.2361e-12],
 [8.4365e-13, 0.0000e+00, 3.9639e-12],
 [3.2982e-23, 0.0000e+00, 2.4506e-24],
 [3.8354e-24, 0.0000e+00, 9.8442e-26],
 [8.6697e-25, 0.0000e+00, 9.0727e-23],

[5.4935e-23, 0.0000e+00, 9.1679e-22],
[1.9377e-27, 5.6052e-45, 2.4616e-26],
[1.7591e-27, 4.2039e-45, 1.5319e-27],
[1.2612e-44, 6.0068e-31, 1.9128e-42],
[0.0000e+00, 3.4832e-11, 0.0000e+00],
[0.0000e+00, 1.4501e-12, 0.0000e+00],
[1.9618e-44, 1.4354e-30, 3.6462e-42],
[0.0000e+00, 6.3410e-25, 0.0000e+00],
[0.0000e+00, 4.0280e-28, 0.0000e+00],
[1.7375e-28, 1.8217e-44, 3.4536e-28],
[1.5585e-30, 1.2752e-43, 1.1295e-33],
[2.4924e-36, 8.3169e-38, 2.6008e-36],
[0.0000e+00, 6.6365e-19, 0.0000e+00],
[9.0515e-41, 5.6939e-34, 8.3796e-40],
[0.0000e+00, 2.0413e-29, 0.0000e+00],
[7.8192e-43, 6.5466e-32, 9.8458e-41],
[1.1242e-26, 0.0000e+00, 2.9358e-30],
[3.1744e-23, 0.0000e+00, 1.3017e-23],
[1.4013e-45, 2.1226e-30, 2.8026e-45],
[1.0022e-40, 6.8044e-34, 2.8821e-39],
[0.0000e+00, 4.9228e-25, 0.0000e+00],
[0.0000e+00, 1.7416e-16, 0.0000e+00],
[2.9427e-44, 3.6251e-32, 5.6052e-45],
[7.9878e-15, 0.0000e+00, 1.3327e-13],
[1.1625e-20, 0.0000e+00, 3.9307e-20],
[1.8572e-04, 0.0000e+00, 9.0569e-07],
[1.0574e-14, 0.0000e+00, 3.3383e-12],
[3.9379e-08, 0.0000e+00, 7.1160e-08],
[8.0254e-12, 0.0000e+00, 3.0634e-11],
[1.7581e-07, 0.0000e+00, 8.8486e-07],
[1.4056e-25, 0.0000e+00, 2.1834e-30],
[7.8917e-01, 0.0000e+00, 8.5079e-01],
[7.1060e-26, 2.8026e-45, 1.4506e-23],
[5.1567e-17, 0.0000e+00, 5.3690e-16],
[1.0573e-12, 0.0000e+00, 7.7589e-11],
[1.5462e-13, 0.0000e+00, 1.3675e-11],
[1.8000e-01, 0.0000e+00, 1.4029e-01],
[4.7314e-17, 0.0000e+00, 8.3416e-16],
[1.8052e-09, 0.0000e+00, 1.7806e-10],
[1.6307e-23, 0.0000e+00, 2.7089e-21],
[1.7611e-02, 0.0000e+00, 3.7681e-03],
[2.0665e-23, 0.0000e+00, 1.6426e-22],
[4.4358e-19, 0.0000e+00, 7.3366e-20],
[2.7721e-13, 0.0000e+00, 4.1531e-12],
[9.6459e-07, 0.0000e+00, 3.3579e-05],
[3.0931e-17, 0.0000e+00, 9.0915e-16],
[5.7089e-05, 0.0000e+00, 9.7161e-05],
[3.6363e-11, 0.0000e+00, 1.8742e-11],
[1.2242e-14, 0.0000e+00, 4.5851e-16],
[2.5199e-26, 2.8026e-45, 1.7108e-24],
[4.3191e-19, 0.0000e+00, 6.0371e-19],
[8.4781e-30, 2.6358e-42, 1.2839e-27],
[2.0008e-16, 0.0000e+00, 6.0068e-16],
[7.0013e-32, 1.7879e-41, 1.7302e-31],
[2.5600e-22, 0.0000e+00, 2.5200e-21],
[3.5621e-35, 5.2125e-38, 6.4667e-34],

[4.1993e-15, 0.0000e+00, 4.0199e-14],
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[5.6101e-23, 0.0000e+00, 7.8495e-24],
[5.5394e-15, 0.0000e+00, 4.4320e-14],
[4.5187e-14, 0.0000e+00, 3.2447e-14],
[1.6883e-12, 0.0000e+00, 2.3961e-11],
[0.0000e+00, 4.5195e-24, 0.0000e+00],
[0.0000e+00, 1.3630e-18, 0.0000e+00],
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[0.0000e+00, 9.3463e-09, 0.0000e+00],
[0.0000e+00, 6.9630e-07, 0.0000e+00],
[0.0000e+00, 3.2838e-19, 0.0000e+00],
[0.0000e+00, 1.0000e+00, 0.0000e+00],
[4.4450e-32, 1.6045e-42, 1.6949e-33],
[1.4994e-43, 1.6817e-31, 8.3938e-42],
[0.0000e+00, 9.2287e-23, 0.0000e+00],
[0.0000e+00, 1.7201e-08, 0.0000e+00],
[5.5722e-37, 1.0618e-36, 7.6952e-36],
[4.0659e-29, 5.6052e-45, 5.3477e-31],
[3.9236e-44, 3.4513e-30, 8.1711e-41],
[2.8026e-45, 2.8086e-30, 1.4714e-43],
[0.0000e+00, 4.0151e-21, 0.0000e+00],
[0.0000e+00, 7.4877e-29, 0.0000e+00],
[4.2677e-39, 1.3897e-36, 2.6798e-40],
[0.0000e+00, 4.4854e-20, 0.0000e+00],
[2.6275e-10, 0.0000e+00, 1.8042e-11],
[3.1204e-23, 0.0000e+00, 5.9996e-23],
[8.7160e-27, 0.0000e+00, 4.2966e-28],
[4.3986e-23, 0.0000e+00, 2.8054e-30],
[1.5008e-38, 1.1654e-37, 2.2841e-43],
[5.8749e-35, 1.1796e-38, 3.6278e-34],
[3.6673e-09, 0.0000e+00, 8.6827e-10],
[1.0955e-10, 0.0000e+00, 4.0628e-12],
[1.2876e-14, 0.0000e+00, 2.5066e-13],
[5.1515e-13, 0.0000e+00, 1.2315e-12],
[8.4270e-11, 0.0000e+00, 6.8147e-10],
[3.1100e-04, 0.0000e+00, 1.3486e-05],
[1.6693e-09, 0.0000e+00, 1.8168e-08],
[1.4817e-22, 0.0000e+00, 5.6804e-21],
[8.1175e-03, 0.0000e+00, 3.2065e-03],
[6.4400e-13, 0.0000e+00, 1.2037e-11],
[2.4635e-03, 0.0000e+00, 1.5596e-03],
[1.6790e-03, 0.0000e+00, 1.4106e-06],
[7.4278e-10, 0.0000e+00, 3.7402e-14],
[5.4094e-05, 0.0000e+00, 3.6979e-07],
[6.8061e-07, 0.0000e+00, 1.1720e-05],
[1.3582e-06, 0.0000e+00, 5.7015e-06],
[2.3529e-19, 0.0000e+00, 5.3890e-23],
[1.7032e-22, 0.0000e+00, 2.2709e-20],
[7.9875e-25, 0.0000e+00, 9.5612e-24],
[2.8129e-14, 0.0000e+00, 1.1477e-14],
[2.5881e-33, 1.9491e-41, 6.4096e-34],
[2.9233e-21, 0.0000e+00, 2.0997e-19],
[4.5517e-12, 0.0000e+00, 1.2104e-16],
[1.0093e-11, 0.0000e+00, 1.6032e-14],
[1.0996e-23, 0.0000e+00, 3.1655e-23],

```
[3.6497e-21, 0.0000e+00, 9.0153e-22],  
[1.2191e-26, 1.4013e-45, 2.4231e-25],  
[1.2054e-22, 0.0000e+00, 1.1762e-22],  
[7.9292e-21, 0.0000e+00, 1.4724e-19],  
[1.8667e-20, 0.0000e+00, 2.4638e-19],  
[2.4848e-27, 2.8026e-45, 6.9955e-28],  
[3.2567e-20, 0.0000e+00, 1.9972e-22],  
[7.9044e-32, 3.3911e-43, 7.1137e-34],  
[5.6993e-32, 6.6982e-43, 4.5927e-34],  
[5.9675e-17, 0.0000e+00, 3.4317e-17],  
[1.6312e-14, 0.0000e+00, 2.8051e-12]])
```

```
In [26]: print(f"Output with softmax activation: {output}")
```

Output with softmax activation: tensor([[6.5089e-38, 3.8771e-13, 2.3017e-19],

[0.0000e+00, 1.2393e-17, 5.6340e-29],
 [2.6289e-35, 1.7317e-10, 9.4923e-19],
 [4.8645e-22, 4.2486e-06, 7.4933e-12],
 [5.3045e-19, 7.3404e-05, 1.6193e-10],
 [2.8268e-25, 1.4714e-08, 6.5907e-13],
 [0.0000e+00, 6.3750e-18, 9.1106e-28],
 [1.9912e-37, 1.2959e-10, 1.0247e-19],
 [0.0000e+00, 1.1955e-13, 1.1266e-24],
 [8.4441e-22, 8.5763e-06, 1.1186e-11],
 [5.0702e-36, 1.9008e-10, 4.4512e-19],
 [2.3588e-29, 1.5912e-10, 1.2513e-14],
 [0.0000e+00, 1.6163e-14, 2.4978e-24],
 [1.1999e-32, 2.8682e-09, 3.1186e-17],
 [3.2684e-33, 4.0591e-10, 1.5594e-17],
 [0.0000e+00, 4.9384e-14, 4.8922e-24],
 [1.2504e-35, 1.2730e-10, 8.7443e-19],
 [8.7785e-23, 2.5844e-06, 2.7044e-12],
 [7.4251e-29, 1.7162e-09, 3.3782e-15],
 [2.4376e-23, 2.2459e-07, 2.6048e-12],
 [0.0000e+00, 1.0317e-16, 5.6071e-30],
 [0.0000e+00, 6.5115e-20, 5.0989e-37],
 [0.0000e+00, 9.1569e-18, 3.4608e-34],
 [0.0000e+00, 2.1678e-13, 1.6265e-24],
 [0.0000e+00, 6.4393e-21, 8.0903e-36],
 [0.0000e+00, 4.4038e-15, 1.7624e-30],
 [0.0000e+00, 1.0461e-18, 2.2935e-34],
 [0.0000e+00, 1.4961e-20, 1.8136e-36],
 [0.0000e+00, 6.5490e-15, 2.1471e-29],
 [0.0000e+00, 2.1018e-14, 2.9226e-30],
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 [0.0000e+00, 1.0491e-20, 5.5666e-33],
 [0.0000e+00, 3.1840e-20, 2.1556e-37],
 [0.0000e+00, 6.7820e-16, 7.4710e-29],
 [0.0000e+00, 1.2800e-19, 3.7905e-37],
 [0.0000e+00, 3.8102e-19, 1.0191e-33],
 [0.0000e+00, 1.8203e-19, 9.2030e-37],
 [0.0000e+00, 4.2703e-20, 5.5814e-38],
 [0.0000e+00, 1.2204e-19, 2.6409e-37],
 [0.0000e+00, 3.2595e-17, 2.0152e-31],
 [0.0000e+00, 1.3519e-20, 2.5040e-38],
 [0.0000e+00, 4.8624e-21, 2.5735e-39],
 [0.0000e+00, 2.4638e-16, 2.5784e-32],
 [0.0000e+00, 2.6572e-18, 7.4231e-34],
 [0.0000e+00, 1.5585e-18, 5.9016e-33],
 [0.0000e+00, 3.3223e-17, 7.7210e-33],
 [0.0000e+00, 3.6921e-18, 5.5578e-33],
 [0.0000e+00, 4.6784e-19, 7.7224e-35],
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 [0.0000e+00, 3.1139e-20, 3.9775e-35],
 [0.0000e+00, 2.3516e-22, 1.3809e-38],
 [0.0000e+00, 4.0438e-19, 3.5425e-36],
 [0.0000e+00, 2.0923e-17, 1.4130e-30],
 [0.0000e+00, 1.5663e-17, 7.6812e-31],
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[0.0000e+00, 3.3538e-16, 7.2052e-31],
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[2.2338e-35, 9.2119e-11, 8.3146e-19],
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[3.3073e-14, 5.1815e-05, 1.5460e-07],
[1.0693e-35, 1.3192e-10, 4.7334e-19],
[2.4734e-28, 3.6776e-08, 2.3030e-15],
[2.6010e-33, 1.2131e-10, 3.3871e-17],
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[0.0000e+00, 1.9140e-16, 2.5187e-26],
[1.4013e-44, 1.3139e-13, 4.6587e-23],
[5.1286e-23, 2.0434e-06, 5.2495e-12],
[1.0050e-39, 6.5690e-12, 1.0943e-20],
[1.1172e-34, 1.1116e-10, 6.3035e-18],
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[0.0000e+00, 3.0169e-17, 4.1561e-30],
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[1.7605e-39, 7.2105e-12, 7.9399e-21],
[1.2771e-28, 4.2135e-08, 8.4507e-16],
[4.5075e-19, 1.6393e-06, 5.9235e-10],
[2.6082e-37, 3.1068e-12, 2.5753e-19],
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[0.0000e+00, 4.8174e-17, 5.0934e-31],
[0.0000e+00, 1.2959e-22, 1.1294e-39],
[0.0000e+00, 1.7069e-17, 3.8438e-34],
[0.0000e+00, 5.8676e-20, 4.0394e-37],
[0.0000e+00, 6.6763e-19, 1.5391e-35],
[0.0000e+00, 2.6320e-20, 7.6154e-38],
[0.0000e+00, 7.3662e-19, 4.2603e-29],
[0.0000e+00, 3.6574e-22, 4.4735e-41],
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[0.0000e+00, 2.8926e-17, 5.4830e-33],
[0.0000e+00, 4.6044e-18, 4.1481e-35],
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[0.0000e+00, 2.9299e-22, 7.0875e-41],
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[0.0000e+00, 6.9674e-21, 4.7940e-39],
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[0.0000e+00, 1.3354e-19, 6.1765e-35],
[0.0000e+00, 9.9458e-15, 7.7321e-29],
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[0.0000e+00, 6.9519e-14, 6.7490e-27],
[0.0000e+00, 3.2795e-18, 8.6196e-34],
[0.0000e+00, 4.3796e-14, 7.5150e-26],
[0.0000e+00, 2.4336e-16, 3.2813e-31],
[0.0000e+00, 1.2797e-12, 1.2897e-24],

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[0.0000e+00, 4.4723e-19, 7.3880e-34],
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[0.0000e+00, 4.1364e-19, 2.4758e-35],
[0.0000e+00, 6.5450e-19, 1.7249e-36],
[2.3491e-28, 2.9547e-08, 4.7320e-15],
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[3.5501e-10, 3.5268e-03, 2.0106e-05],
[4.8470e-07, 1.9295e-02, 2.8058e-04],
[2.0323e-22, 2.1333e-07, 7.0950e-12],
[1.0000e+00, 9.7285e-01, 9.9968e-01],
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[9.4568e-27, 1.9372e-07, 2.4987e-14],
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[1.4433e-43, 9.8489e-13, 9.6241e-23],
[0.0000e+00, 1.1084e-16, 4.2294e-27],
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```

Below are the outputs without softmax activation:

```
In [15]: print(f"\n Output without softmax activation:{output_no_softmax}")
```

Output without softmax activation:tensor([[-84.1552, 267.8842, -149.7228],

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```

```
In [28]: print(f"\n Output without softmax activation:{output_no_softmax}")
```

Output without softmax activation:tensor([[192.1592, 168.8958, -274.339
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```

```
In [32]: print(f"\n Output without softmax activation:{output_no_softmax}")
```

Output without softmax activation:tensor([[261.2544, -245.4799, 140.1274],

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```
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[ 133.8861, -132.9262,  70.1051],
[ 123.9493, -120.6476,  65.2891]])
```

3b

```
In [18]: def train_and_val(model, train_X, train_y, epochs, draw_curve = True):
        """
        Parameters
        -----
        model: a PyTorch model
        train_X: np.array shape(ndata,nfeatures)
        train_y: np.array shape(ndata)
        epochs: int
        draw_curve: bool
        """

        ### Define your loss function, optimizer. Convert data to torch tensor ###
        optimizer = torch.optim.Adam(model.parameters(), lr = 0.005)
        loss_func = nn.CrossEntropyLoss()
        train_X = torch.tensor(train_X, dtype=torch.float)
        train_y = torch.tensor(train_y, dtype=torch.long)

        ### Split training examples further into training and validation ###
        X_train, X_val, y_train, y_val = train_test_split(train_X, train_y)

        val_array=[]
        lowest_val_loss = np.inf
        model_param = model.state_dict()

        for i in range(epochs):
            ### Compute the loss and do backpropagation ###
            optimizer.zero_grad()
            y_pred = model(X_train)
            loss = loss_func(y_pred, y_train-1)
            loss.backward()
            optimizer.step()

            ### compute validation loss and keep track of the lowest val loss ###
            with torch.no_grad():
                val_pred = model(X_val)
                val_loss = loss_func(val_pred, y_val-1).detach().numpy()
                val_array.append(val_loss)

            if val_loss < lowest_val_loss:
                lowest_val_loss = val_loss
                model_param = model.state_dict()
```

```

# The final number of epochs is when the minimum error in validation se
final_epochs=np.argmin(val_array)+1
print("Number of epochs with lowest validation:",final_epochs)
### Recover the model weight ###
model.load_state_dict(model_param)

if draw_curve:
    plt.figure()
    plt.plot(np.arange(len(val_array))+1,val_array,label='Validation los
    plt.xlabel('Epochs')
    plt.ylabel('Loss')
    plt.legend()

```

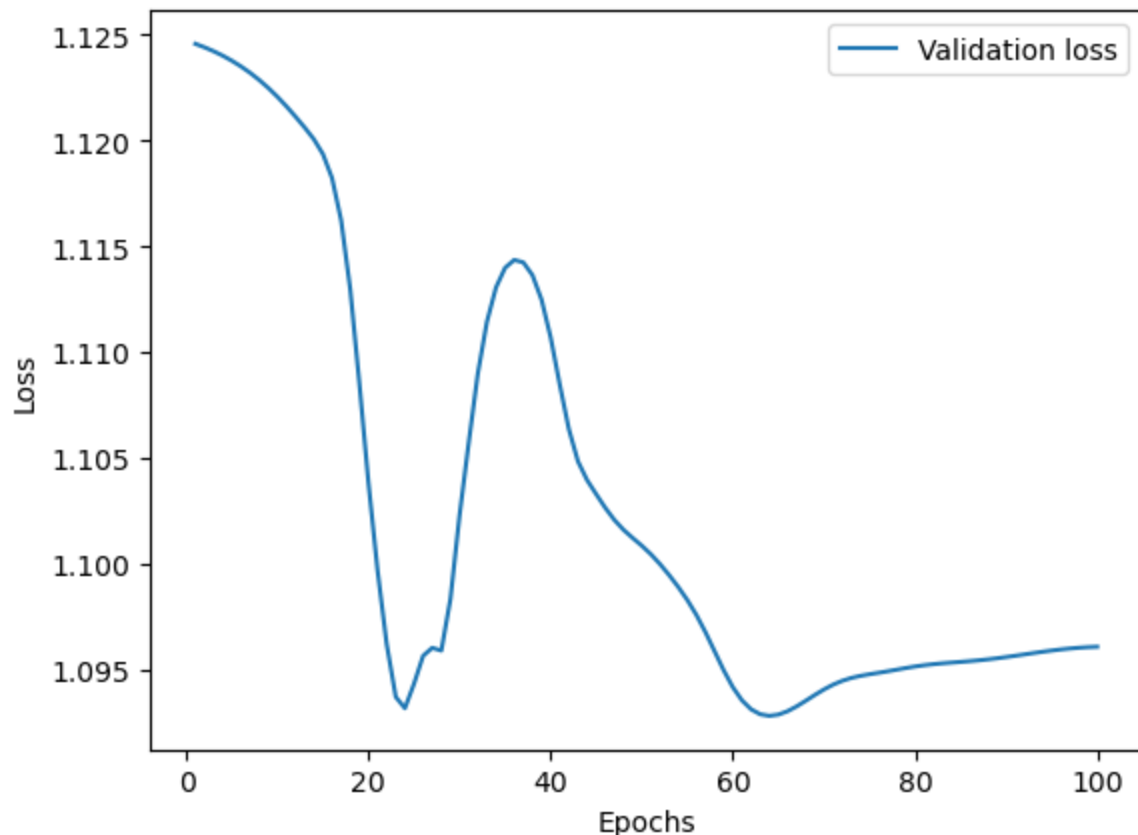
```
In [19]: train_and_val(model_w_softmax, wines_train_X, wines_train_y, 100)
```

Number of epochs with lowest validation: 64

```

/var/folders/m8/skfw9g2x4_g4pq5cv80_g24w0000gn/T/ipykernel_8594/2244072975.p
y:14: UserWarning: To copy construct from a tensor, it is recommended to use
sourceTensor.clone().detach() or sourceTensor.clone().detach().requires_grad
_(True), rather than torch.tensor(sourceTensor).
    train_X = torch.tensor(train_X, dtype=torch.float)
/var/folders/m8/skfw9g2x4_g4pq5cv80_g24w0000gn/T/ipykernel_8594/2244072975.p
y:15: UserWarning: To copy construct from a tensor, it is recommended to use
sourceTensor.clone().detach() or sourceTensor.clone().detach().requires_grad
_(True), rather than torch.tensor(sourceTensor).
    train_y = torch.tensor(train_y, dtype=torch.long)

```

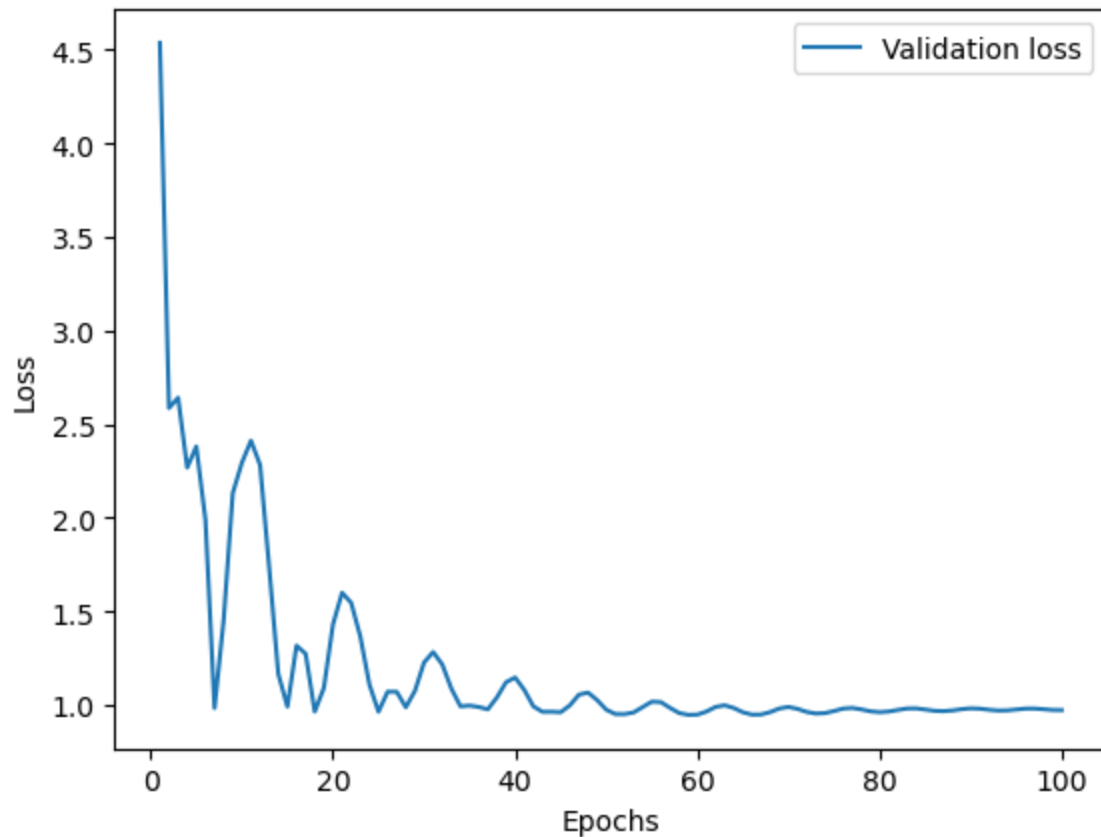


The loss starts to increase again at ~25, but the lowest validation is at 64

```
In [36]: train_and_val(model_wo_softmax, wines_train_X, wines_train_y, 100)
```

Number of epochs with lowest validation: 59

```
/var/folders/m8/skfw9g2x4_g4pq5cv80_g24w0000gn/T/ipykernel_8594/2244072975.p
y:14: UserWarning: To copy construct from a tensor, it is recommended to use
sourceTensor.clone().detach() or sourceTensor.clone().detach().requires_grad
_(True), rather than torch.tensor(sourceTensor).
train_X = torch.tensor(train_X, dtype=torch.float)
/var/folders/m8/skfw9g2x4_g4pq5cv80_g24w0000gn/T/ipykernel_8594/2244072975.p
y:15: UserWarning: To copy construct from a tensor, it is recommended to use
sourceTensor.clone().detach() or sourceTensor.clone().detach().requires_grad
_(True), rather than torch.tensor(sourceTensor).
train_y = torch.tensor(train_y, dtype=torch.long)
```



```
In [ ]: ### CALCULATE ACCURACY OF EACH
# calculate_accuracy(model,xs,ys)
def calculate_accuracy_nn(model,xs,ys):
    y_pred=np.zeros_like(ys)
    for idx,x in enumerate(xs):
        y_pred[idx]=model.forward(x)
    return np.sum(ys==y_pred)/len(ys)

calculate_accuracy_nn(model_w_softmax, wines_train_X, wines_train_y)
calculate_accuracy_nn(model_wo_softmax, wines_train_X, wines_train_y)
```