## HW3Q3

```
In [1]:
        import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        from sklearn.model selection import train test split
        from sklearn.preprocessing import StandardScaler
        from sklearn.neighbors import KNeighborsClassifier
        from sklearn.naive_bayes import GaussianNB
        from sklearn.linear model import LogisticRegression
        from sklearn.metrics import accuracy score
        from sklearn.decomposition import PCA
In [2]: path = 'data/marriage.csv'
        df = pd.read_csv(path, header = None)
        X = df.iloc[:,:-1]
        y = df.iloc[:,-1:].values.ravel()
        seed = 55
        X_train, X_test, y_train, y_test = train_test_split(X, y, test]
        Q3.1
        #naive bayes
In [3]:
        nb clf = GaussianNB(var_smoothing=1e-3 ).fit(X_train, y_train)
        nb pred = nb clf.predict(X test)
        nb_acc = accuracy_score(y_test, nb_pred)
In [4]: #logistic regression
        lr clf = LogisticRegression().fit(X train, y train)
        lr pred = lr clf.predict(X test)
        lr_acc = accuracy_score(y test, lr_pred)
        #knn
In [5]:
        knn_clf = KNeighborsClassifier().fit(X_train, y_train)
        knn_pred = knn_clf.predict(X_test)
        knn acc = accuracy score(y test, knn pred)
        print("NB Testing Accuracy:", nb_acc)
In [6]:
        print("LR Testing Accuracy:", lr_acc)
        print("KNN Accuracy:", knn_acc)
```

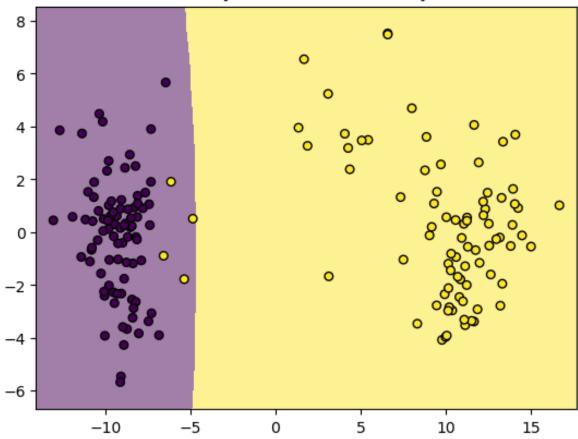
NB Testing Accuracy: 0.9705882352941176 LR Testing Accuracy: 0.9705882352941176

KNN Accuracy: 0.9705882352941176

## Q3.2

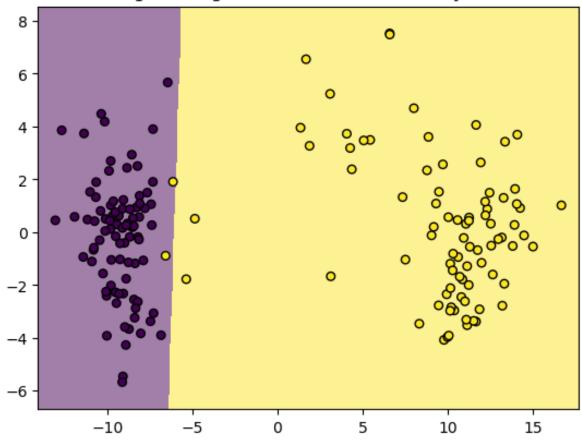
```
#PCA on the data
 In [7]:
          pca = PCA(n components=2)
          X fit = pca.fit transform(X)
          X train, X test, y train, y test = train test split(X fit, y, 1
          nb clf pca= GaussianNB().fit(X train, y train)
          lr clf pca = LogisticRegression().fit(X train, y train)
          knn_clf_pca = KNeighborsClassifier().fit(X_train, y_train)
          x \min, x \max = X \text{ fit}[:, 0].\min() - 1, X \text{ fit}[:, 0].\max() + 1
          y_{min}, y_{max} = X_{fit}[:, 1].min() - 1, <math>X_{fit}[:, 1].max() + 1
          xx, yy = np.meshgrid(np.arange(x min, x max, 0.05), np.arange(\sqrt{}
          nb_Z = nb_clf_pca.predict(np.c_[xx.ravel(), yy.ravel()])
          nb Z = nb Z.reshape(xx.shape)
          lr Z = lr_clf pca.predict(np.c [xx.ravel(), yy.ravel()])
          lr Z = lr Z.reshape(xx.shape)
          knn Z = knn_clf_pca.predict(np.c_[xx.ravel(), yy.ravel()])
          knn Z = knn Z.reshape(xx.shape)
         #plot nb
In [15]:
          plt.pcolormesh(xx, yy, nb_Z, alpha=0.5, cmap='viridis')
          plt.scatter(X_fit[:, 0], X_fit[:, 1], c=y, s=30, edgecolor='k')
          plt.title("Naive Bayes Decision Boundary Plot")
          plt.savefig("q3plot1")
```

## Naive Bayes Decision Boundary Plot



```
In [16]: #plot logistic regression
   plt.pcolormesh(xx, yy, lr_Z, alpha=0.5, cmap='viridis')
   plt.scatter(X_fit[:, 0], X_fit[:, 1], c=y, s=30, edgecolor='k')
   plt.title("Logistic Regression Decision Boundary Plot")
   plt.savefig("q3plot2")
```

## Logistic Regression Decision Boundary Plot



```
In [17]: #plot knn
plt.pcolormesh(xx, yy, knn_Z, alpha=0.5, cmap='viridis')
plt.scatter(X_fit[:, 0], X_fit[:, 1], c=y, s=30, edgecolor='k')
plt.title("K-Nearest Neighbors Decision Boundary Plot")
plt.savefig("q3plot3")
```

K-Nearest Neighbors Decision Boundary Plot

