```
In [54]: import numpy as np
   import pandas as pd
   import matplotlib.pyplot as plt
   import statsmodels.api as sm
   import statistics
   from sklearn.linear_model import LinearRegression
   from sklearn.linear_model import LogisticRegression
   from sklearn.discriminant_analysis import QuadraticDiscriminantAnalysis
   from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
   from sklearn.neighbors import KNeighborsClassifier
   from sklearn.metrics import confusion_matrix
   from sklearn.model_selection import train_test_split
   from sklearn.model_selection import LeaveOneOut
```

### Problem a

```
In [55]: # normal distribution mean and standard deviation
    mu = 0
    sigma = 1

# X: 100 random samples from normal distribution N(mu = 0, std = 1)
    x = np.random.normal(mu, sigma, 100)

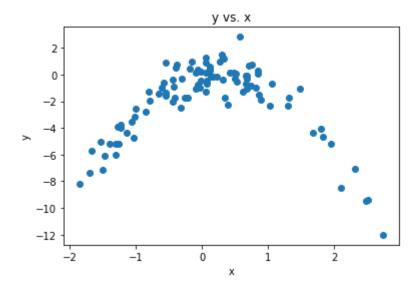
# noise: 100 random samples from normal distribution N(mu = 0, std = 1)
    noise = np.random.normal(mu, sigma, 100)

# Response: y = x - 2*x^2 + noise
    y = x - 2*x**2 + noise
```

## **Problem b**

```
In [56]: # Scatterplot of x and y
    plt.scatter(x, y)
    plt.title("y vs. x")
    plt.xlabel("x")
    plt.ylabel("y")
```

Out[56]: Text(0,0.5,'y')



# Problem c

```
In [67]:
         # ----- MODEL 1 -----
         # Make dataset of desired predictors
         predictors = {
                 'x': x,
         data = pd.DataFrame(predictors)
         data = np.asarray(data)
         # ----- LOOCV -----
         loo = LeaveOneOut()
         n = loo.get_n_splits(data)
         sum MSE = 0
         for train index, test index in loo.split(data):
             # Obtain training and testing data splits
             X_train, X_test = data[train_index], data[test_index]
             y_train, y_test = y[train_index], y[test_index]
             # Create inear Regression Model and fit data
             LinRegr classifier = LinearRegression()
             LinRegr_classifier.fit(X_train, y_train)
             # Make prediction on test data
             y_pred = LinRegr_classifier.predict(X_test)
             # Calculate MSE
             current mse = (y \text{ test}[0] - y \text{ pred}[0])**2
             sum MSE = sum MSE + current mse
         avg mse = sum MSE/n
         print("The LOOCV error from Model 1 is " + repr(avg_mse))
```

The LOOCV error from Model 1 is 8.427931437291429

```
In [58]:
          ----- MODEL 2 -----
         # Make dataset of desired predictors
         predictors = {
                 'x': x,
                'x^2': x**2,
         data = pd.DataFrame(predictors)
         data = np.asarray(data)
         data = sm.add constant(data)
         # ----- LOOCV -----
         loo = LeaveOneOut()
         n = loo.get_n_splits(data)
         sum MSE = 0
         for train_index, test_index in loo.split(data):
            # Obtain training and testing data splits
            X_train, X_test = data[train_index], data[test_index]
            y_train, y_test = y[train_index], y[test_index]
            # Create inear Regression Model and fit data
            LinRegr_classifier = LinearRegression()
            LinRegr_classifier.fit(X_train, y_train)
            # Make prediction on test data
            y pred = LinRegr classifier.predict(X test)
            # Calculate MSE
            current_mse = (y_test[0] - y_pred[0])**2
            sum MSE = sum MSE + current mse
         avg mse = sum MSE/n
         print("The LOOCV error from Model 2 is " + repr(avg_mse))
```

The LOOCV error from Model 2 is 0.882104115348292

```
In [59]:
             ----- MODEL 3 -----
         # Make dataset of desired predictors
         predictors = {
                 'x': x,
                 'x^2': x**2,
                 x^3: x^{*3}
         data = pd.DataFrame(predictors)
         data = np.asarray(data)
         # ----- LOOCV -----
         loo = LeaveOneOut()
         n = loo.get_n_splits(data)
         sum\ MSE = 0
         for train_index, test_index in loo.split(data):
             # Obtain training and testing data splits
             X_train, X_test = data[train_index], data[test_index]
             y_train, y_test = y[train_index], y[test_index]
             # Create inear Regression Model and fit data
             LinRegr_classifier = LinearRegression()
             LinRegr_classifier.fit(X_train, y_train)
             # Make prediction on test data
             y pred = LinRegr classifier.predict(X test)
             # Calculate MSE
             current mse = (y \text{ test}[0] - y \text{ pred}[0])**2
             sum MSE = sum MSE + current mse
         avg mse = sum MSE/n
         print("The LOOCV error from Model 3 is " + repr(avg_mse))
```

The LOOCV error from Model 3 is 0.894180442253765

```
In [60]:
          ----- MODEL 4 -----
         # Make dataset of desired predictors
         predictors = {
                 'x': x,
                'x^2': x**2,
                x^3: x^{*3}
                 x^4: x^{*4},
               }
         data = pd.DataFrame(predictors)
         data = np.asarray(data)
         # ----- LOOCV -----
         loo = LeaveOneOut()
         n = loo.get n splits(data)
         sum_MSE = 0
         for train_index, test_index in loo.split(data):
            # Obtain training and testing data splits
            X_train, X_test = data[train_index], data[test_index]
            y train, y test = y[train index], y[test index]
            # Create inear Regression Model and fit data
            LinRegr classifier = LinearRegression()
            LinRegr_classifier.fit(X_train, y_train)
            # Make prediction on test data
            y pred = LinRegr classifier.predict(X test)
            # Calculate MSE
            current_mse = (y_test[0] - y_pred[0])**2
            sum MSE = sum MSE + current mse
         avg_mse = sum_MSE/n
         print("The LOOCV error from Model 4 is " + repr(avg_mse))
```

The LOOCV error from Model 4 is 0.9063418943218408

# Problem d

```
In [68]:
         # ----- MODEL 1 -----
         # Make dataset of desired predictors
         predictors = {
                 'x': x,
         data = pd.DataFrame(predictors)
         data = np.asarray(data)
         data = sm.add_constant(data)
         # Create the Multiple Linear Regression Model and fit it
         MLRmodel = sm.OLS(y, data)
         result = MLRmodel.fit()
         # Create table of model prediction results
         data = {'Coefficient Beta_i': result.params,
                 't-Values': result.tvalues,
                 'p-Values': result.pvalues
         data analysis = pd.DataFrame(data)
         data_analysis.round(4) # Round values in table to 4-decimal places
```

### Out[68]:

	Coefficient Beta_i	t-Values	p-Values
0	-1.8842	-6.6805	0.0000
1	-0.0785	-0.2781	0.7815

```
In [70]:
           ----- MODEL 2 -----
         # Make dataset of desired predictors
         predictors = {
                 'x': x,
                 'x^2': x**2,
         data = pd.DataFrame(predictors)
         data = np.asarray(data)
         data = sm.add_constant(data)
         # Create the Multiple Linear Regression Model and fit it
         MLRmodel = sm.OLS(y, data)
         result = MLRmodel.fit()
         # Create table of model prediction results
         data = {'Coefficient Beta_i': result.params,
                 't-Values': result.tvalues,
                 'p-Values': result.pvalues
         data_analysis = pd.DataFrame(data)
         data_analysis.round(4) # Round values in table to 4-decimal places
```

### Out[70]:

	Coefficient Beta_i	t-Values	p-Values
0	-0.0511	-0.4506	0.6533
1	1.0246	10.1348	0.0000
2	-1.9310	-28.3073	0.0000

```
In [71]:
             ----- MODEL 3 -----
         # Make dataset of desired predictors
         predictors = {
                 'x': x,
                 'x^2': x**2,
                 'x^3': x**3,
         data = pd.DataFrame(predictors)
         data = np.asarray(data)
         data = sm.add_constant(data)
         # Create the Multiple Linear Regression Model and fit it
         MLRmodel = sm.OLS(y, data)
         result = MLRmodel.fit()
         # Create table of model prediction results
         data = {'Coefficient Beta_i': result.params,
                 't-Values': result.tvalues,
                 'p-Values': result.pvalues
         data_analysis = pd.DataFrame(data)
         data_analysis.round(4) # Round values in table to 4-decimal places
```

### Out[71]:

	Coefficient Beta_i	t-Values	p-Values
0	-0.0651	-0.5319	0.5960
1	1.0730	5.8018	0.0000
2	-1.9083	-19.1667	0.0000
3	-0.0195	-0.3136	0.7545

```
In [72]:
             ----- MODEL 4 -----
         # Make dataset of desired predictors
         predictors = {
                 'x': x,
                 'x^2': x**2,
                 x^3: x^{**3}
                 'x^4': x**4,
         data = pd.DataFrame(predictors)
         data = np.asarray(data)
         data = sm.add_constant(data)
         # Create the Multiple Linear Regression Model and fit it
         MLRmodel = sm.OLS(y, data)
         result = MLRmodel.fit()
         # Create table of model prediction results
         data = {'Coefficient Beta_i': result.params,
                 't-Values': result.tvalues,
                 'p-Values': result.pvalues
         data analysis = pd.DataFrame(data)
         data_analysis.round(4) # Round values in table to 4-decimal places
```

### Out[72]:

	Coefficient Beta_i	t-Values	p-Values
0	-0.0403	-0.3081	0.7587
1	1.1433	5.0937	0.0000
2	-2.0016	-10.2573	0.0000
3	-0.0650	-0.6319	0.5290
4	0.0290	0.5565	0.5792

```
In [ ]:
```