Group 5

The first step of our EDA is exploring the shape of our data (6819, 96). We followed up by looking at all the 96 columns of data and checking for missing values. There are none. We then isolated the columns with two or fewer unique values, which were 'Bankrupt?', 'Liability-Assets Flag,' and 'Net Income Flag.' We determined that 'Bankrupt?' would be the dependent variable for our models. From the data, we calculated only 3.3% of companies were bankrupted, which shows an imbalance in the response variable. We observed 220 cases of bankruptcy (1) and 6,599 cases of non-bankruptcy (0), making it a relatively rare occurrence. We removed 'Net Income Flag' because it was uniform (every value was 1) and didn't provide any predictive value. We also removed 'Liability -Assets Flag' because only eight rows had values of 1 (the rest had 0), so it also would not provide a good predictive value for our models.

Our data preparation continued with additional feature selection. However prior to performing feature selection, we removed features that were showing collinearity from the dataset by calculating the VIF for each feature and removing any with a VIF value greater than 5. After dropping these features, we were left with 64 columns. Then we used SelectKBest from sklearn on these 64 features for our feature selection, selecting the 10 dependent variables the model determined to be the best predictors of the 'Bankrupt?' variable. These were ROA(C) before interest and depreciation before interest, Tax rate (A), Net Value Per Share (B), Operating Profit Per Share (Yuan ¥), Net worth/Assets, Retained Earnings to Total Assets, Total expense/Assets, Equity to Long-term Liability, Current Liability to Current Assets and Net Income to Stockholder's Equity.

Next, we used an isolation forest to remove 5% of the anomalies within the dataset using IsolationForest from sklearn. The isolation forest looked for anomalies across all potential independent (predictor) variables. After removing 5% of the rows with anomalies, the shape of our dataframe shrunk (6,478, 10). We split the remaining data into two sets: training (80%; 5,182 values) and testing (20%; 1,296 values). We followed by scaling the independent

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variables for both datasets using StandardScalar from sklearn. Finally, we used SMOTE (Synthetic Minority Over-Sampling Technique) from imblearn to create an unbalanced number of 'Bankrupt?' rows in our training data. Oversampling from our training data allows us to better train our model for rare events, such as bankruptcy.

Using the k-fold cross-validation method, we built three machine learning models to test our dataset. We found that the naïve Bayes model best classified our testing value with an F1 score of 0.28. Our logistic regression model did similarly well classifying our testing data and received a marginally lower F1 score of 0.26. Both models also performed closely during their goodness of fit tests, receiving TPR (Recall) values of 0.82 and 0.80, FPR values of 0.13 and 0.14, precision values of 0.17 and 0.16, and accuracy values of 0.87 and 0.86. In comparison, our SVM had an F1 score of 0.11 after hyperparameter tuning, making it our worst performing model. It performed similarly to the previous two models on the goodness of fit tests with the exception of recall, where it got a score of 0.08. Hyperparameter tuning consists of using GridSearchCV to find the best hyperparameters for our model, which end up being C=10 and gamma=1. Accuracy for all three models ranged from 0.86 to 0.96; however, we expect high accuracy as most of the predicted values will be 0 (no bankruptcy). To see the complete results for each model, see our appendix.

We believe our models are overfitting the training data. We believe overfitting is plaguing our models because of the large drop in the goodness of fit tests between our training and testing sets. We tried to address this in several ways, including implementing SMOTE, stratifying our sample selection to account for representative portions of bankrupt properties, and k-folds. These changes only marginally improved our F scores, though we kept them where they made sense. Additional areas for exploration include the impact of isolation forests on the data (perhaps only removing non-bankrupt anomalies) and alternative methods of scaling and data selection.

Module 4 Assignment 2 Module 4 Assignment 2: Company Bankruptcy Prediction

Group 5 Scott Jue Zach Watson

Appendix:

| | Model | TPR | FPR | precision | recall | accuracy | f1-value |
|---|---------------------|--------------|--------------|--------------|--------------|--------------|----------|
| 0 | SVM | [0.99, 0.08] | [0.92, 0.01] | [0.97, 0.18] | [0.99, 0.08] | [0.96, 0.96] | 0.11 |
| 1 | Logistic Regression | [0.86, 0.8] | [0.2, 0.14] | [0.99, 0.16] | [0.86, 0.8] | [0.86, 0.86] | 0.26 |
| 2 | naive bayes | [0.87, 0.82] | [0.18, 0.13] | [0.99, 0.17] | [0.87, 0.82] | [0.87, 0.87] | 0.28 |
| | | | | | | | |

Intro

Links

https://canvas.northwestern.edu/courses/167719/assignments/1078603?module_item_id=2319248

https://www.kaggle.com/datasets/fedesoriano/company-bankruptcy-prediction

Modules

```
In [1]:
         #For data manipulation and visualization
         import pandas as pd
         import seaborn as sns
         import matplotlib.pyplot as plt
         import numpy as np
         from numpy import array
         from numpy import arange
         #For Isolation Forest from sklearn
         from sklearn.ensemble import IsolationForest
         from enum import auto
         #From sklearn (SVM, Logistic, Bayes)
         from sklearn.svm import SVC
         from sklearn import svm
         from sklearn.linear model import LogisticRegression
         from sklearn.linear_model import ElasticNet
         from sklearn.linear model import ElasticNetCV
         from sklearn.model selection import GridSearchCV
         from sklearn.model selection import RepeatedKFold
         from sklearn.model selection import train test split
         from sklearn.model selection import KFold
         from sklearn.model selection import cross val score
         from sklearn.feature_selection import SelectKBest
         from sklearn.feature selection import f classif
         from sklearn.preprocessing import MinMaxScaler, StandardScaler
         from sklearn.naive bayes import GaussianNB
         from sklearn import metrics
         from sklearn.metrics import f1_score, classification_report, confusion_matrix
         from sklearn.metrics import roc curve, auc, roc auc score
         from sklearn.metrics import PrecisionRecallDisplay
         from sklearn.metrics import precision score, recall score, precision recall curve
         from sklearn.metrics import mean squared error
         #Other
         from math import sqrt
```

Import Files

```
In [2]: #Import data.csv from the Kaggle page linked above
    # from google.colab import files
    # uploaded = files.upload()
In [3]: df = pd.read_csv("data.csv")
```

EDA

Intro Stats

```
In [4]:
         df.shape
        (6819, 96)
Out[4]:
In [5]:
         df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 6819 entries, 0 to 6818
        Data columns (total 96 columns):
             Column
         #
                                                                         Non-Null Count Dtype
             -----
                                                                         -----
         0
             Bankrupt?
                                                                         6819 non-null
                                                                                          int64
              ROA(C) before interest and depreciation before interest
         1
                                                                         6819 non-null
                                                                                          float64
         2
               ROA(A) before interest and % after tax
                                                                         6819 non-null
                                                                                          float64
               ROA(B) before interest and depreciation after tax
                                                                         6819 non-null
         3
                                                                                          float64
         4
               Operating Gross Margin
                                                                         6819 non-null
                                                                                          float64
               Realized Sales Gross Margin
         5
                                                                         6819 non-null
                                                                                          float64
                                                                                          float64
         6
               Operating Profit Rate
                                                                         6819 non-null
               Pre-tax net Interest Rate
         7
                                                                         6819 non-null
                                                                                          float64
               After-tax net Interest Rate
                                                                         6819 non-null
                                                                                          float64
         8
         9
               Non-industry income and expenditure/revenue
                                                                         6819 non-null
                                                                                          float64
         10
              Continuous interest rate (after tax)
                                                                         6819 non-null
                                                                                          float64
                                                                                          float64
         11
              Operating Expense Rate
                                                                         6819 non-null
         12
               Research and development expense rate
                                                                         6819 non-null
                                                                                          float64
         13
               Cash flow rate
                                                                         6819 non-null
                                                                                          float64
         14
              Interest-bearing debt interest rate
                                                                         6819 non-null
                                                                                          float64
                                                                                          float64
         15
              Tax rate (A)
                                                                         6819 non-null
         16
              Net Value Per Share (B)
                                                                         6819 non-null
                                                                                          float64
         17
              Net Value Per Share (A)
                                                                         6819 non-null
                                                                                          float64
              Net Value Per Share (C)
         18
                                                                         6819 non-null
                                                                                          float64
         19
              Persistent EPS in the Last Four Seasons
                                                                         6819 non-null
                                                                                          float64
         20
              Cash Flow Per Share
                                                                                          float64
                                                                         6819 non-null
              Revenue Per Share (Yuan ¥)
                                                                                          float64
         21
                                                                         6819 non-null
                                                                                          float64
         22
               Operating Profit Per Share (Yuan ¥)
                                                                         6819 non-null
         23
               Per Share Net profit before tax (Yuan ¥)
                                                                         6819 non-null
                                                                                          float64
               Realized Sales Gross Profit Growth Rate
                                                                                          float64
         24
                                                                         6819 non-null
         25
              Operating Profit Growth Rate
                                                                         6819 non-null
                                                                                          float64
              After-tax Net Profit Growth Rate
                                                                         6819 non-null
                                                                                          float64
         26
         27
               Regular Net Profit Growth Rate
                                                                         6819 non-null
                                                                                          float64
         28
               Continuous Net Profit Growth Rate
                                                                         6819 non-null
                                                                                          float64
         29
              Total Asset Growth Rate
                                                                                          float64
                                                                         6819 non-null
         30
              Net Value Growth Rate
                                                                         6819 non-null
                                                                                          float64
               Total Asset Return Growth Rate Ratio
                                                                         6819 non-null
                                                                                          float64
```

| | Group_o_modulo_+_, toolgriment_2 | | |
|----------|--|--------------------------------|--------------------|
| 32 | Cash Reinvestment % | 6819 non-null | float64 |
| 33 | Current Ratio | 6819 non-null | float64 |
| 34 | Quick Ratio | 6819 non-null | float64 |
| 35 | Interest Expense Ratio | 6819 non-null | float64 |
| 36 | Total debt/Total net worth | 6819 non-null | float64 |
| 37 | Debt ratio % | 6819 non-null | float64 |
| 38 | Net worth/Assets | 6819 non-null | float64 |
| 39 | Long-term fund suitability ratio (A) | 6819 non-null | float64 |
| 40 | Borrowing dependency | 6819 non-null | float64 |
| 41 | Contingent liabilities/Net worth | 6819 non-null | float64 |
| 42 | Operating profit/Paid-in capital | 6819 non-null | float64 |
| 43 | Net profit before tax/Paid-in capital | 6819 non-null | float64 |
| 44 | Inventory and accounts receivable/Net value | 6819 non-null | float64 |
| 45 | Total Asset Turnover | 6819 non-null | float64 |
| 46 | Accounts Receivable Turnover | 6819 non-null | float64 |
| 47 | Average Collection Days | 6819 non-null | float64 |
| 48 | Inventory Turnover Rate (times) | 6819 non-null | float64 |
| 49 | Fixed Assets Turnover Frequency | 6819 non-null | float64 |
| 50 | Net Worth Turnover Rate (times) | 6819 non-null | float64 |
| 51 | Revenue per person | 6819 non-null | float64 |
| 52 53 | Operating profit per person | 6819 non-null 6819 non-null | float64 float64 |
| 53 54 | Allocation rate per person Working Capital to Total Assets | 6819 non-null | float64 |
| 55 | Quick Assets/Total Assets | 6819 non-null | float64 |
| 56 | Current Assets/Total Assets | 6819 non-null | float64 |
| 57 | Cash/Total Assets | 6819 non-null | float64 |
| 58 | Quick Assets/Current Liability | 6819 non-null | float64 |
| 59 | Cash/Current Liability | 6819 non-null | float64 |
| 60 | Current Liability to Assets | 6819 non-null | float64 |
| 61 | Operating Funds to Liability | 6819 non-null | float64 |
| 62 | Inventory/Working Capital | 6819 non-null | float64 |
| 63 | Inventory/Current Liability | 6819 non-null | float64 |
| 64 | Current Liabilities/Liability | 6819 non-null | float64 |
| 65 | Working Capital/Equity | 6819 non-null | float64 |
| 66 | Current Liabilities/Equity | 6819 non-null | float64 |
| 67 | Long-term Liability to Current Assets | 6819 non-null | float64 |
| 68 | Retained Earnings to Total Assets | 6819 non-null | float64 |
| 69 | Total income/Total expense | 6819 non-null | float64 |
| 70 | Total expense/Assets | 6819 non-null | float64 |
| 71 | Current Asset Turnover Rate | 6819 non-null | float64 |
| 72 | Quick Asset Turnover Rate | 6819 non-null | float64 |
| 73 | Working capitcal Turnover Rate | 6819 non-null | float64 |
| 74 | Cash Turnover Rate | 6819 non-null | float64 |
| 75 | Cash Flow to Sales | 6819 non-null | float64 |
| 76 | Fixed Assets to Assets | 6819 non-null | float64 |
| 77 | Current Liability to Liability | 6819 non-null | float64 |
| 78 | Current Liability to Equity | 6819 non-null | float64 |
| 79 | Equity to Long-term Liability | 6819 non-null | float64 |
| 80 | Cash Flow to Total Assets | 6819 non-null | float64 |
| 81 | Cash Flow to Liability | 6819 non-null | float64 |
| 82 | CFO to Assets | 6819 non-null | float64 |
| 83 | Cash Flow to Equity | 6819 non-null | float64 |
| 84 | Current Liability to Current Assets | 6819 non-null | float64 |
| 85 | Liability-Assets Flag | 6819 non-null | int64 |
| 86 | Net Income to Total Assets | 6819 non-null | float64 |
| 87 | Total assets to GNP price | 6819 non-null | float64 |
| 88 | No-credit Interval | 6819 non-null | float64 |
| 89 | Gross Profit to Sales | 6819 non-null | float64 |
| 90 | Net Income to Stockholder's Equity | 6819 non-null | float64 |
| 91 | Liability to Equity | 6819 non-null | float64 |
| 92 | Degree of Financial Leverage (DFL) | 6819 non-null | float64 |
| 93 | Interest Coverage Ratio (Interest expense to EBIT) | 6819 non-null | float64 |
| 94 | Net Income Flag | 6819 non-null | int64 |
| 95 | Equity to Liability | 6819 non-null | float64 |

dtypes: float64(93), int64(3)
memory usage: 5.0 MB

```
In [6]: # check for missing values
    print(df.isna().sum().sum())
    print(np.isnan(df).sum().sum())
    print(df.isnull().sum().sum())

0
0
0
0
df.head(10)
```

Out[7]: ROA(C) ROA(A) ROA(B) before before Realized After-Pre-tax before Operating Operating interest and interest **Sales** net tax net **Bankrupt?** interest and Gross **Profit** depreciation and % Gross Interest Interest depreciation Margin Rate before after Margin Rate Rate after tax interest tax 0 1 0.370594 0.424389 0.405750 0.601457 0.601457 0.998969 0.796887 0.808809 1 1 0.464291 0.538214 0.516730 0.610235 0.610235 0.998946 0.797380 0.809301 2 1 0.426071 0.499019 0.472295 0.601450 0.601364 0.998857 0.796403 0.808388 3 1 0.399844 0.451265 0.457733 0.583541 0.583541 0.998700 0.796967 0.808966 0.465022 0.538432 0.522298 0.598783 0.598783 0.998973 0.797366 0.809304 1 5 1 0.388680 0.415177 0.419134 0.590171 0.590251 0.998758 0.796903 0.808771 0 6 0.390923 0.445704 0.436158 0.619950 0.619950 0.998993 0.797012 0.808960 7 0 0.508361 0.570922 0.559077 0.601738 0.601717 0.999009 0.797449 0.809362 8 0 0.488519 0.545137 0.543284 0.603612 0.603612 0.998961 0.797414 0.809338

0.542963

0.599209 0.599209

0.999001 0.797404 0.809320

10 rows × 96 columns

0

9

```
In [8]: # create a for loop to get the categorical columns with 2 or less than 2 unique values
list_1=[]
for i in df.columns:
    x=df[i].value_counts()
    if len(x)<=2:
        list_1.append(i)
    else:
        continue</pre>
In [9]: # categorical variables (value_counts <= 2 )
list_1
```

Out[9]: ['Bankrupt?', ' Liability-Assets Flag', ' Net Income Flag']

0.495686 0.550916

Checking [Bankrupt?]

```
In [10]:
           df['Bankrupt?'].unique()
Out[10]: array([1, 0], dtype=int64)
In [11]:
           df['Bankrupt?'].value_counts()
                6599
Out[11]:
                220
          Name: Bankrupt?, dtype: int64
In [12]:
           sns.histplot(x="Bankrupt?", data=df)
Out[12]: <AxesSubplot:xlabel='Bankrupt?', ylabel='Count'>
            6000
            5000
            4000
            3000
            2000
            1000
                  0.0
                           0.2
                                     0.4
                                              0.6
                                                       0.8
                                                                1.0
                                       Bankrupt?
```

Checking [Net Income Flag]

```
In [13]: df[' Net Income Flag'].unique()
Out[13]: array([1], dtype=int64)
In [14]: df[' Net Income Flag'].value_counts()
Out[14]: 1    6819
    Name: Net Income Flag, dtype: int64
    Every value is a 1 for this feature, so we can drop this column since it doesn't provide us any predictive value.
```

Checking [Liability-Assets Flag]

```
In [15]: df[' Liability-Assets Flag'].unique()
Out[15]: array([0, 1], dtype=int64)
```

```
In [16]: df[' Liability-Assets Flag'].value_counts()
```

Out[16]: 0 6811 1 8

Name: Liability-Assets Flag, dtype: int64

This feature may not provide good predictive value to the model since it has a large imbalance between the 0 and 1 classes. Therefore, we can also drop this column.

Feature Selection

```
In [17]:
            y = df['Bankrupt?']
In [18]:
            X = df.drop(columns = ['Bankrupt?',' Liability-Assets Flag',' Net Income Flag'])
In [19]:
            X.head()
Out[19]:
                    ROA(C)
                              ROA(A)
                                            ROA(B)
                    before
                              before
                                                                 Realized
                                                                                        Pre-tax
                                                                                                    After-
                                            before
                                                     Operating
                                                                           Operating
                                                                                                                  Nor
               interest and
                                                                    Sales
                              interest
                                                                                            net
                                                                                                  tax net
                                       interest and
                                                         Gross
                                                                               Profit
                                                                                                                    in
               depreciation
                               and %
                                                                    Gross
                                                                                        Interest
                                                                                                  Interest
                                       depreciation
                                                        Margin
                                                                                Rate
                                                                                                           expenditur<sub>(</sub>
                    before
                                after
                                                                  Margin
                                                                                           Rate
                                                                                                     Rate
                                           after tax
                   interest
                                  tax
           0
                  0.370594 0.424389
                                           0.405750
                                                                0.601457
                                                                                                 0.808809
                                                      0.601457
                                                                            0.998969
                                                                                       0.796887
           1
                  0.464291 0.538214
                                           0.516730
                                                      0.610235  0.610235
                                                                            0.998946
                                                                                       0.797380
                                                                                                 0.809301
           2
                  0.426071 0.499019
                                           0.472295
                                                      0.601450 0.601364
                                                                                                 0.808388
                                                                            0.998857
                                                                                       0.796403
           3
                  0.399844 0.451265
                                                      0.583541
                                           0.457733
                                                                0.583541
                                                                            0.998700
                                                                                       0.796967
                                                                                                 0.808966
           4
                  0.465022 0.538432
                                           0.522298
                                                      0.598783 0.598783
                                                                            0.998973
                                                                                      0.797366
                                                                                                0.809304
           5 rows × 93 columns
In [20]:
            cor matrix = X.corr().abs()
            cor_matrix
                               ROA(C)
                                         ROA(A)
Out[20]:
                                                       ROA(B)
                                before
                                          before
                                                                             Realized
                                                                                                    Pre-tax
                                                                                                               After-
                                                                                       Operating
                                                        before
                                                                 Operating
                                                                                Sales
                           interest and
                                         interest
                                                                                                       net
                                                                                                              tax net
                                                   interest and
                                                                     Gross
                                                                                           Profit
                                           and %
                          depreciation
                                                                               Gross
                                                                                                   Interest
                                                                                                             Interest
                                                   depreciation
                                                                    Margin
                                                                                            Rate
                                before
                                            after
                                                                              Margin
                                                                                                      Rate
                                                                                                                 Rate
                                                      after tax
                               interest
                                             tax
```

| | ROA(C) before interest and depreciation before interest | ROA(A) before interest and % after tax | ROA(B) before interest and depreciation after tax | Operating Gross Margin | Realized Sales Gross Margin | Operating Profit Rate | Pre-tax net Interest Rate | After- tax net Interest Rate |
|--|--|---|---|------------------------------|--------------------------------------|-----------------------------|------------------------------------|---------------------------------------|
| ROA(C) before interest and depreciation before interest | 1.000000 | 0.940124 | 0.986849 | 0.334719 | 0.332755 | 0.035725 | 0.053419 | 0.049222 |
| ROA(A) before interest and % after tax | 0.940124 | 1.000000 | 0.955741 | 0.326969 | 0.324956 | 0.032053 | 0.053518 | 0.049474 |
| ROA(B) before interest and depreciation after tax | 0.986849 | 0.955741 | 1.000000 | 0.333749 | 0.331755 | 0.035212 | 0.053726 | 0.049952 |
| Operating Gross Margin | 0.334719 | 0.326969 | 0.333749 | 1.000000 | 0.999518 | 0.005745 | 0.032493 | 0.027175 |
| Realized Sales Gross Margin | 0.332755 | 0.324956 | 0.331755 | 0.999518 | 1.000000 | 0.005610 | 0.032232 | 0.026851 |
| ••• | | ••• | | | | | ••• | |
| Net Income to Stockholder's Equity | 0.274287 | 0.291744 | 0.280617 | 0.075304 | 0.074891 | 0.006216 | 0.011343 | 0.010648 |
| Liability to Equity | 0.143629 | 0.141039 | 0.142838 | 0.085434 | 0.085407 | 0.001541 | 0.004043 | 0.004390 |
| Degree of Financial Leverage (DFL) | 0.016575 | 0.011515 | 0.014663 | 0.011806 | 0.011268 | 0.000935 | 0.000855 | 0.000927 |
| Interest Coverage Ratio (Interest expense to EBIT) | 0.010573 | 0.013372 | 0.011473 | 0.001167 | 0.001158 | 0.000393 | 0.000984 | 0.000957 |
| Equity to Liability | 0.052416 | 0.057887 | 0.056430 | 0.120029 | 0.120196 | 0.017071 | 0.014559 | 0.010900 |

93 rows × 93 columns

From the correlation matrix, tt appears that there are feature variables that are highly correlated

with one another. We will remove these features from the data set before perform any feature selection

```
In [21]:
          # create function that calculates VIF and drops any columns that have a VIF greater tha
          from numpy.core.fromnumeric import shape
          from statsmodels.stats.outliers influence import variance inflation factor
          def cal vif(x):
            thresh = 5
            output = pd.DataFrame()
            k= x.shape[1]
            vif = [variance_inflation_factor(x.values, i) for i in range(x.shape[1])]
            for i in range(1, k):
              print('Iteration no ', i)
              print(vif)
              a = np.argmax(vif)
              print ('Max vif is for variable no : ',a)
              if(vif[a]<=thresh):</pre>
                break
              if(i==1):
                output = x.drop(x.columns[a], axis = 1)
                vif = [variance_inflation_factor(output.values, j) for j in range(output.shape[1]
              elif(i>1):
                output = output.drop(output.columns[a], axis = 1)
                vif = [variance inflation factor(output.values, j) for j in range(output.shape[1]
            return(output)
```

In [22]:

```
# select non-correlated features
select_features = cal_vif(X)
```

Iteration no 1 [69.01850223325629, 43.595467498059946, 102.89448309616877, 63784107.00172185, 1076.2927 070171597, 82724488.6699668, 31113694.835407037, 1822.4631663986613, 11440456.186908506, 337.99634964330147, 1.2612409105505844, 1.092355471002177, 7.558642941248357, 1.04160101 64740737, 1.3274514511808797, 782.0303535509482, 3911.8118072742327, 3126.1592381455343, 32.038789339109286, 4.187190702735464, 4.360882460291321, 637.9946060235118, 21.71372243 8768315, 1.0395640475297192, 1.9329780110347787, 133.49730672423615, 132.68280523255828, 1.023089394986167, 1.1253183182354223, 1.1245861424549934, 1.826570517281875, 3.64806105 38571487, 1.3580387765690056, 1.0598223919041465, 1.0089769111560956, 1.557035661827639 3, 2537479393.75135, 1151884255.8721192, 1.4719169862979808, 37.479577314825995, 9.11908 4282253652, 638.7817489939426, 30.099870538396818, 36.079332620209954, 5.90068744505700 2, 1.568334527356616, 1.3076566055626773, 1.1411513437741085, 1.3401406515691616, 6.3814 957656262425, 3.3370896492705473, 1.5984649281962882, 1.151680847434043, 8312138253.9460 25, 4.863768733631312, 8967464419.896431, 3.1145919262637225, 1.2531060287816265, 1.0655 728847958652, 2114448416.2389877, 8.623759081912636, 1.0899417821964243, 1.0741889852962 132, 434920292358.3289, 42.996715926290285, 9683442458.85514, 1.029557800052723, 3.79869 9602643246, 1.3418897111441632, 3.3378088704393223, 1.5157152200219561, 1.44058475309982 64, 86.72381556431932, 1.151810963594114, 26.558956154340503, 1.2166252388920598, 160501 777557.3512, 1986153299.4930298, 44.95782251671022, 4.633900338678031, 2.81804221958423 7, 7.4803865390129625, 3.2764813532019925, 3.099771595988936, 29.082462082248224, 1.0916 79437424625, 2620.003074424558, 52974519.67115891, 8.349087282557573, 464.9450812993252 4, 1.004899088279558, 1.0069812599816976, 3.0450084439958767] Max vif is for variable no : 63 Iteration no 2 [69.01849691369395, 43.59547048292976, 102.8944865314102, 45578056.325474106, 1076.29083 49953073, 68308332.20316012, 46068881.32398531, 1822.4576227516136, 7136073.092476168, 3 37.99511970725786, 1.261240912451517, 1.0923554695906437, 7.558645384225325, 1.041601019 320039, 1.3274514539783513, 782.0301752237611, 3911.8108874884506, 3126.1588800846826, 3 2.03878883363618, 4.187190706401702, 4.360881713945248, 637.9943461281914, 21.7137215477 01944, 1.0395640461995121, 1.9329778931122417, 133.49732377366854, 132.68279090823086, 1.0230893946861304, 1.1253182946254912, 1.1245861462000974, 1.8265709508961552, 3.648061 118436004, 1.3580387738872863, 1.0598224170899029, 1.0089769097717793, 1.557035362706019 8, 1211921977.3012173, 1246209950.0808964, 1.4719169448010654, 37.47957704835284, 9.1190 84481804702, 638.7817381667278, 30.0998743148141, 36.07932331548993, 5.900687470576614, 1.5683339437828248, 1.3076566061989865, 1.1411513439350474, 1.340140640292022, 6.3814958 66761502, 3.337073719766715, 1.5984650371176408, 1.1516808567785228, 6174376701.131676, 4.863768742509252, 13176896329.1313, 3.1145919377996094, 1.2531060320096226, 1.065572880 3702742, 3787709310.6710753, 8.623760019443726, 1.089941866044864, 1.0741889774828752, 4 2.99671643059254, 3809215778.499866, 1.0295578084596269, 3.7986996206177217, 1.341889710 0604439, 3.337808803397488, 1.5157152199107773, 1.4405847531006957, 86.72353249226317, 1.1518109644499996, 26.558954474115097, 1.2166252323527484, 8.619078994517915, 125756174 8.9007561, 44.95781981933953, 4.633900332086082, 2.818041498472827, 7.480386547328878, 85798949, 32781497.65832567, 8.349082271787305, 464.9450602206384, 1.004899087309812, 1. 0069812600425747, 3.0450082921478168]

Max vif is for variable no : 55

Iteration no 3

[69.01849452372532, 43.595470267106144, 102.8944888446852, 63526829.67938712, 1076.28956 84104674, 67449322.50493494, 36668056.32584292, 1822.4639003690547, 13962690.489861302, 337.99573244484674, 1.2612409139579566, 1.092355470891468, 7.558644609569187, 1.04160101 3294724, 1.3274514634520032, 782.0300177438577, 3911.825952981428, 3126.1641226330944, 3 2.03878886176478, 4.187190703805333, 4.360882547422468, 637.9943904369215, 21.7137224525 05955, 1.0395640470138046, 1.93297801954716, 133.4973633149779, 132.68282884770704, 1.02 30893978140831, 1.1253183218243321, 1.1245861455582309, 1.8265709289296184, 3.6480611377 014553, 1.358038777343423, 1.0598224210093286, 1.008976915117549, 1.5570356017044513, 95 757525.73492745, 35471693.21959462, 1.4719170222464582, 37.47957934088521, 9.11908350442 9282, 638.7816091840853, 30.099868096862995, 36.07932201100475, 5.900687643949427, 1.568 334507096248, 1.3076566066955773, 1.14115132902465, 1.3401406537378822, 6.38149631000269 85, 3.3370938322465635, 1.5984649052784057, 1.1516808641123286, 6.850010363090277, 4.863 768796571675, 3.114591608609848, 1.2531060197421915, 1.0655728113977028, 33.718849639342 636, 8.62375864304305, 1.0899418243434618, 1.0741889664733162, 42.996712953108954, 28359 54732.3820934, 1.0295578042764377, 3.7986997364766046, 1.3418897086913806, 3.33780887200 39956, 1.5157152126424858, 1.4405847481025786, 86.72252362201279, 1.151810960420001, 26. 558951208266095, 1.2166252443327839, 8.619078154576739, 279969841.3454677, 44.9578218185 6884, 4.633900372795803, 2.8180408075072747, 7.480386604741033, 3.2764811675186443, 3.09 97716201987635, 29.08243457587624, 1.0916794238140572, 1.015994196304741, 35683127.14422 739, 8.34909029781619, 464.9450554611016, 1.0048990883202875, 1.0069812721710625, 3.0450 1064519454]

Max vif is for variable no : 63

Iteration no 4

[69.01849416878743, 43.59547129817059, 102.89447808824264, 56333875.92703587, 1076.29066 9948118, 97354701.53565718, 41532461.7135313, 1822.4609084693732, 6637921.079015497, 33 7.995151580485, 1.2612409031825165, 1.0923554765379158, 7.55864413925379, 1.041601014374 4194, 1.327451456345836, 782.0306460293407, 3911.827075596561, 3126.167972099581, 32.038 78951547213, 4.1871906720760155, 4.360906717210306, 637.9945313941229, 21.71372243173617 6, 1.0395640467403182, 1.9329777760305522, 133.49731870034776, 132.68281795657464, 1.023 089397586892, 1.1253183566271887, 1.1245861462346474, 1.826570619302632, 3.6480610980882 484, 1.358038777090801, 1.0598223181460762, 1.0089768985927776, 1.5570356384119086, 1716 42295.05580112, 160632429.69394273, 1.471916931811809, 37.47958075391307, 9.119084042957 78, 638.7818101608594, 30.09987231068251, 36.07932717804895, 5.900687639026617, 1.568333 1391096824, 1.307656606109157, 1.1411513341678727, 1.3401406461448464, 6.38149589452587 9, 3.3370799219583294, 1.5984650284629915, 1.1516808367327809, 6.850009916264347, 4.8637 68047863697, 3.114591662786962, 1.2531060128658835, 1.0655729014445212, 33.7188468727548 9, 8.623760503526261, 1.0899418376841932, 1.0741889966380838, 42.99671657014475, 1.02955 7804758303, 3.798699681012043, 1.341889710514457, 3.3378087891026316, 1.515715213917648 2, 1.4405847568354493, 86.72306663342752, 1.1518109620617498, 26.558947152803444, 1.2166 252497559482, 8.619079736381892, 292.1031351697963, 44.95782145390332, 4.63389936571168 2, 2.8180413259742267, 7.480386577295594, 3.276481126808994, 3.099771547682144, 29.08246 814531047, 1.0916794107504266, 1.0159941834938921, 47258441.1994593, 8.349090458384103, 464.9450499451531, 1.0048990884392115, 1.0069812417982584, 3.0449961472066387

Max vif is for variable no : 36

Iteration no 5

[69.01847762393106, 43.59545629125267, 102.89448898452184, 52333058.66342763, 1076.29105 7065112, 341.86853522587774, 2099.5261498713908, 1822.4156372873792, 89.10478775436573,

337.99688012573296, 1.2612409737848675, 1.0923554858548772, 7.558638598482653, 1.0416009 502780854, 1.3274515150105766, 782.0321094108765, 3911.823467111887, 3126.1853163104315, 32.03878771160532, 4.187190720013486, 4.3608926766893195, 637.9945685710077, 21.71372368 094246, 1.039564045095158, 1.9329775421141169, 133.49736959611602, 132.68282126897205, 1.0230894320420947, 1.1253182783212856, 1.124586104565759, 1.8265709189241315, 3.6480608 302272017, 1.3580387726756127, 1.0598222947276759, 1.0089769466746221, 1.557035158922801 3, 27.90003513018217, 1.4719167916608338, 37.47957857063784, 9.119081462277464, 638.7817 152302136, 30.099869082602744, 36.079122241338034, 5.9006892036747915, 1.56833384791112 2, 1.3076566137276333, 1.1411512362171545, 1.3401406769641084, 6.381497853757042, 3.3370 90973766631, 1.598464557736682, 1.151680828058747, 6.850009417979535, 4.863769816416459, 3.1145815221288236, 1.2531059465957277, 1.0655729200764783, 33.71883769246864, 8.6237487 0253754, 1.08994175891492, 1.074188815422581, 42.996606938151416, 1.0295577898792077, 3. 798699994342887, 1.3418897017748495, 3.337808872282332, 1.5157151902615524, 1.4405847323 109686, 86.72312217134055, 1.9669767399687135, 26.558933905110095, 1.2166252207918875, 8.619078805331862, 292.1031931048324, 44.95774248236059, 4.63389477365156, 2.81803932473 67386, 7.480386700728255, 3.276474040047572, 3.099771563142134, 29.082466281481675, 1.09 16794591181704, 1.0159942838802982, 43829817.44859567, 8.349081144384643, 464.9443469237 583, 1.004899084961915, 1.006981380446472, 3.0450012847882006] Max vif is for variable no : 3

Iteration no 6

[69.01843867910554, 43.59545155672999, 102.8944102797821, 1076.2428663838002, 341.867660 0541442, 2099.5112510482295, 1822.432122978828, 89.10801894192626, 337.99679759847317, 1.261241136214421, 1.0923559147560635, 7.558636455709332, 1.0416009967834616, 1.32745136 05677942, 782.0324549132838, 3911.821395485279, 3126.1843379482093, 32.03876085247057, 4.1871928234659395, 4.3608694560476255, 637.9944893905695, 21.713698273370486, 1.0395640 239409107, 1.9329782325964255, 133.4973699620581, 132.68280169131896, 1.023089393909463 9, 1.1253170607912437, 1.1245860240422372, 1.8265711137156035, 3.648059398345722, 1.3580 381762903415, 1.0598228725557348, 1.0089769775690551, 1.55703726894736, 27.9000601537756 66, 1.4719165524901863, 37.479578523191776, 9.119082640748614, 638.781778152115, 30.0998 6594671374, 36.07910227807162, 5.900690100875294, 1.5683337255431786, 1.307657423274896, 1.1411508012554488, 1.3401415388440343, 6.381495643813511, 3.3371155156563144, 1.5984657 131466147, 1.1516811618248535, 6.850033627600692, 4.863767970788844, 3.11459933682107, 1.2531059002883342, 1.0655730197354143, 33.71893126219203, 8.623754050878121, 1.08994167 17803085, 1.0741889930741357, 42.99662148073233, 1.0295582500737832, 3.7987046642321687, 1.3418899038773715, 3.3378204841297334, 1.5157150902984617, 1.4405845608547672, 86.72209 005391389, 1.9669772181726175, 26.558982723869715, 1.2166251760216373, 8.61910407274050 7, 292.1036490320659, 44.957807354953296, 4.633893835101491, 2.8180456230496027, 7.48038 2231327735, 3.276475997415841, 3.0997723952860623, 29.08245863777649, 1.091679396725050 2, 1.0159939615629514, 1078.073456111489, 8.349086619171208, 464.9452748796325, 1.004899 069268558, 1.0069814817649314, 3.0450037874832705]

Max vif is for variable no : 15

Iteration no 7

[69.01148842280259, 43.59543521273421, 102.87553947178513, 1076.2446883692207, 341.85259 26976999, 2099.5172715558, 1822.4414028564318, 89.10205557786338, 337.96915800890457, 1. 260743772250644, 1.0923551438771026, 7.5585593129109405, 1.041505641947889, 1.3274326845 370847, 625.9314921082365, 627.3011997478857, 31.961787177022206, 4.187177922100799, 4.3 60675007174375, 636.1790553280997, 21.7136569776252, 1.0395633907822985, 1.9327616248915 893, 133.4973100434927, 132.6827872638896, 1.0230864954618184, 1.125075001003408, 1.1245 227136372413, 1.825212159708601, 3.647927730629745, 1.3580297147721954, 1.05899507461525 34, 1.008974509274567, 1.5570215510717864, 27.900050687332882, 1.4719118602277756, 37.47 772848138426, 9.11907719666441, 637.1070167663637, 30.065426089794993, 36.0731512219868 3, 5.900681717786368, 1.5682017179055168, 1.3076536389120572, 1.1411453218209113, 1.3401 318292972995, 6.381096210416007, 3.337103653571233, 1.5983852964568894, 1.15168116857138 26, 6.849779535022583, 4.863410116264991, 3.114286491986602, 1.2531053273130492, 1.06017 67815577352, 33.71744480405118, 8.623742914311695, 1.0899244765596845, 1.074171236595820 5, 42.99118403316912, 1.0294503080709454, 3.7909345606316607, 1.341883656399145, 3.33695 54625851667, 1.5156684336974175, 1.4403989725519113, 86.72327558294819, 1.96648603462644 3, 26.558343788519345, 1.216590530153356, 8.618798534232416, 291.93780350501953, 44.9305 0396374942, 4.633893779395728, 2.8180110378080463, 7.480152165180083, 3.276096968973808, 3.099716187128777, 29.081577383758294, 1.0916389436334015, 1.015968756131698, 1078.07393 1115598, 8.34587046245242, 464.5290433862755, 1.0048937888718084, 1.0069728705209218, 3. 0449112590933596]

Max vif is for variable no : 5

Iteration no 8

[68.4134734485659, 43.50972212672066, 102.26028805121341, 1061.4051760459301, 112.648927

93385806, 320.9750261043557, 20.55456430961513, 324.5451884837412, 1.2604069116862935, 1.092354407416861, 7.441135715626964, 1.0413139697900025, 1.3273330260209502, 625.913569 0291458, 627.2835138115104, 31.947674673173267, 4.187120336306441, 3.440443159376746, 63 6.1733725950429, 21.711527490581023, 1.0395610528743506, 1.9176657625480673, 133.4884085 1898368, 132.68235765504684, 1.023066268640549, 1.123926114863505, 1.1244039958736716, 1.8251757015491883, 3.6466770704806484, 1.3579656705773613, 1.0563196280089766, 1.008160 577571725, 1.5422944080737702, 27.849378926875012, 1.4673482481737667, 37.47607948444843 4, 9.113587994545734, 637.0940201941198, 30.048253536460063, 34.610647929066076, 5.87914 1484406465, 1.5636971290823176, 1.307653260322935, 1.140574743760343, 1.340051289710868, 6.374943560666061, 3.0478377805275345, 1.5956497529160363, 1.1512580486676558, 6.8180992 58265109, 4.856338596314535, 3.111267122040018, 1.2530465819443362, 1.0595722463106942, 33.71114405778911, 8.54649305362933, 1.086178967973029, 1.07195984240462, 41.33865791631 088, 1.0293210944258242, 3.789487912771961, 1.3418269452452112, 3.336276371299808, 1.515 6679426105317, 1.4397549897423902, 34.22732653956349, 1.9664653863434682, 26.17436973889 5027, 1.2160512727402109, 8.614818968360677, 288.7515049588338, 44.907585645793276, 4.59 4445279412396, 2.771155667319219, 7.4769560233315895, 3.240025192080528, 3.0995937045092 36, 29.072807693898916, 1.0910175321876878, 1.0159675548144578, 1062.372263166342, 8.309 310435690316, 464.5267743553508, 1.00489356697638, 1.006374734339555, 2.918589041638095 5]

Max vif is for variable no : 80

Iteration no 9

[68.39449609043916, 43.489822762874425, 102.25851367288533, 2.4245019553252987, 112.6166 8484903899, 320.12000193091797, 20.554523175033488, 324.04183196478425, 1.26039184033136 8, 1.0922793832998863, 7.439639212429367, 1.0412906721982869, 1.3272555355909548, 625.88 83883648263, 627.2691219791429, 31.933715591824026, 4.186610873974044, 3.439015615913936 7, 636.1452568174832, 21.705971271149096, 1.0395274457211026, 1.9173763751741684, 133.48 836653763786, 132.68233610758452, 1.023060716926921, 1.1233539925854101, 1.1243898006347 193, 1.8251605616278601, 3.64536905243727, 1.3577891200354013, 1.0563040454039605, 1.008 095588381134, 1.542289124181311, 27.825610381740475, 1.4671772410083435, 37.474648141060 335, 9.113478244084284, 637.0916402315853, 30.043003812765132, 34.603979381599586, 5.878 945179804318, 1.5636809223981092, 1.3075744905884077, 1.1394610848676894, 1.339957907457 9542, 6.373687536799511, 3.0435157408863174, 1.5953389201452504, 1.1510776432585712, 6.8 17227452084542, 4.856084058078039, 3.108873678919052, 1.2530411759743194, 1.059501492740 904, 33.66408934982373, 8.545034396741992, 1.0861728261626475, 1.0719171561173453, 41.32 950432131996, 1.0291547261258673, 3.788649927627811, 1.3418254911217244, 3.330299793746 2, 1.515664007215092, 1.439747497443798, 34.194294358276586, 1.9659653451667396, 26.1743 58795537337, 1.2160364934240178, 8.588486876242918, 288.7244519753587, 44.8948456302564 9, 4.594122199110429, 2.770518214071917, 7.475521425740334, 3.2400216367215067, 3.098460 2399135475, 29.051858784311026, 1.091007322243369, 1.0159345960679802, 8.30886702738564 9, 464.3326112623763, 1.0046552658170353, 1.0063728841247264, 2.918032691797213]

Max vif is for variable no : 37

Iteration no 10

[68.07428602945784, 43.01754208962404, 100.93568450007014, 2.424501667477437, 112.573527 9084649, 320.1174160774162, 20.55043366312831, 323.98774273541113, 1.260369397800261, 1. 707511702570952, 7.439356103976971, 1.0390089040243449, 1.326995420710958, 625.801819313 0976, 627.1890381542659, 30.279887853090983, 4.185905527891765, 3.4389544894291117, 6.57 9045095312046, 21.59010861002325, 1.0395266229874272, 1.8742187905100824, 133.4880795468 61, 132.68204318267607, 1.0230434601513054, 1.1232703577991623, 1.1243570778959862, 1.41 46241889203581, 3.6453681343422497, 1.356975505154817, 1.056263730869551, 1.008092265352 924, 1.5422847664896155, 27.823664379458705, 1.4671484349296144, 37.420239884895494, 9.1 06796963428263, 26.644069718749638, 34.60287877260406, 5.87863410170114, 1.5633113385425 619, 1.3075089700251608, 1.139426507993833, 1.3399210971640154, 6.373568945925476, 3.043 4638492285733, 1.594838153091664, 1.1510765869049693, 6.81372530159987, 4.85411189988385 7, 3.1077067621187657, 1.2530327880481387, 1.0595002404635372, 33.663911138938, 8.544688 246348386, 1.0861386596757805, 1.0718921658754632, 41.32922520500351, 1.028717347420751 8, 3.7886349815137597, 1.3416233921050609, 3.327716078183802, 1.5156041694837823, 1.4396 902935794815, 34.19373781902523, 1.9659370327371373, 26.17415218125515, 1.21602900565235 7, 8.587552700416845, 288.7175188761239, 44.88877390139976, 4.53888356203088, 2.76829629 92396867, 7.47552065643155, 3.138511970247934, 3.092184248840577, 29.041082554209627, 1. 0909939274773663, 2619.758076364628, 8.307629157695919, 464.2730636830461, 1.00465109191 81137, 1.0063643224268382, 2.917915966566564]

Max vif is for variable no : 78

Iteration no 11

[68.0671708751222, 3161.0145433695734, 100.9290350607289, 3134.1205350917417, 111.168092 9406005, 320.03445542007415, 20.34226382816748, 323.4744145844528, 1.2603266015382024,

1.0919620955193658, 7.439186231811552, 1.062958577726563, 2.239648037447612, 21033.55865 5599234, 20972.427304388315, 30.2750771901795, 4.185602249253705, 3.4403365179270575, 6. 579015980033503, 688.1931398669997, 4.617288581547319, 11660.246326760267, 133.487442979 95515, 132.681410242651, 1.0230428993045766, 1.123123223924885, 1.1245428860119775, 106 5.7758417024213, 1225.8275681857187, 1.356704982059388, 1.0574095975908262, 1.0080921076 124225, 1.5432730696634762, 27.823042855627733, 1.609942952988234, 37.41550995194915, 9. 102402813200193, 965.3561427332245, 34.601585980078525, 5.8757578359138956, 1.5665535543 02068, 1.3094285581886307, 1.1394255184317164, 1.3399060103854485, 13.42526454430752, 3. 0388048787404407, 240.77103544305453, 1.1527582067970257, 1301.841159834393, 23.89348018 501388, 5.575797326518829, 1.2533763974354695, 1.0651050409227334, 33.66292862492502, 8. 543344823757291, 1.0861319093448913, 1.0817467646750039, 164144.96763492495, 1.037992811 1189547, 3.7886165183575278, 1.4007211089378964, 3.3276815547813294, 1.5151710675085264, 1.439252823336188, 34.193281588958214, 1.1510476300828114, 26.1702321797811, 1.216205923 4204619, 8.587187852500454, 288.71038329195636, 44.88629980005726, 4.538641476735129, 66 1.4109507206955, 775.1977318431635, 1863.9995603715747, 6.3160293891496035, 11678.789845 856334, 1.093665667223448, 8.303850987692048, 464.26714912477377, 4.109120394339819, 1.0 063627640389299, 5.556445161809211]

Max vif is for variable no : 57

Iteration no 12

[68.06364698817885, 3160.9989115896287, 100.88813564037439, 3130.3678198024268, 53.26353 190206734, 256.94231907370875, 12.779158303245312, 321.6756258393183, 1.259297807670863 6, 1.0911253558391656, 7.413554105288364, 1.0611238888017134, 2.2360668964547736, 625.44 62388127977, 626.8031762423668, 30.270999809897226, 4.1810901866800725, 3.40777956270242 3, 6.5585604301258265, 687.2069065312554, 1.0394447308224952, 11659.927182666444, 133.48 32492076588, 132.6797469024261, 1.0230428938973282, 1.1206945109679765, 1.12364753264852 65, 1063.6685608712905, 1202.1806624637154, 1.3549828728293933, 1.0572797127722842, 1.00 78330478175284, 1.5410110116681806, 27.734048075099928, 1.608953631333057, 37.4003310520 4169, 10.687801342080034, 963.8773265711395, 8430.037631771069, 5.766039451825087, 1.560 1769293531467, 1.3092913254457281, 1.1393784622477339, 1.561801466097746, 13.07941098151 3687, 3.0117049142092074, 240.76473117299963, 1.1524742704415225, 1258.9040427952527, 2 1.509315736937577, 5.2777179199154745, 1.220898173794392, 1.0649981555915389, 33.6404318 439812, 872.7949492116599, 1.085112411259928, 1.0808139542118673, 1.0378499748916983, 3. 775278342779325, 1.3409469812889054, 3.3147692456088675, 1.5144888357051538, 1.436387251 06447, 31.820638491012083, 1.1509721786880822, 26.16870957172786, 1.2162059209660576, 8. 497940970193282, 252.15109486382318, 44.61217532908272, 836.7802767798829, 660.679925016 1213, 770.5566061109564, 1545.9998817432102, 6.273450943837458, 11672.589579712996, 1.09 35885841617397, 8.112888257570434, 462.6755947440985, 4.109006009581564, 1.0063493899188 436, 5.515242677330463]

Max vif is for variable no : 75

Iteration no 13

[66.43139361028616, 1750.803660329809, 99.83907795111206, 3010.3937407150265, 53.2241181 463765, 256.9172587150092, 12.713126899425626, 321.6597223600774, 1.247917826372504, 1.0 909675930146927, 7.331948430393472, 1.0605082250446771, 2.2232421169411705, 625.25406252 17648, 626.349916612107, 30.25943696064527, 4.173839142658836, 3.3988996848194493, 6.556 06864099194, 21.548984064688202, 1.0389148552518102, 11658.867030312744, 133.45997333452 542, 132.6636784396683, 1.0230172944353344, 1.0932503269504308, 1.123646495151337, 1060. 6918170765875, 1186.1398058055418, 1.354981952922246, 1.057234588203875, 1.0077946823463 43, 1.536444809201701, 27.21859674375319, 1.6082784481221408, 37.36767091789081, 10.4778 6605079272, 963.8557119677059, 8251.252743988329, 16.985962454642483, 1.558819180510688 8, 1.306664618269137, 1.1393291637306457, 1.5615863229883062, 13.078904737710829, 3.0041 41721294353, 240.6842178635953, 1.1520980947331139, 1255.9192637926308, 21.5056219847331 07, 5.273564952471922, 1.2181115609616462, 1.0649297647741218, 33.35877350888881, 871.59 39324232014, 1.0851078947163642, 1.080777714928296, 1.036635881115165, 3.293050466883853 4, 1.3399146442577885, 2.758303570691326, 1.5144700978097252, 1.434767270380947, 31.7258 49178249824, 1.1489333332059115, 25.98174300988828, 1.2161701437178638, 8.45709112916959 4, 251.05585452895343, 44.429643641945965, 836.7529456706098, 660.4616836537919, 756.137 5003625011, 2.5987442794663314, 6.273428978901807, 1.0891416107248775, 7.62195236167420 7, 456.2267082304405, 4.108935849434929, 1.006313368992015, 5.514368522839727]

Max vif is for variable no : 21

Iteration no 14

[66.39970833523162, 1750.7780383259433, 99.7781672297016, 2.3279529799937952, 52.2076109 1303561, 253.60821117224984, 12.629923997321434, 320.88735919389865, 1.2471599254612933, 1.0909651365344526, 7.327054822686703, 1.0605051878314147, 2.222992613038069, 625.250477 3435616, 626.3481523240276, 30.25943777125143, 4.124878243533241, 3.3987421316340796, 6. 555953750670392, 21.451563716188353, 1.03879484871023, 132.69673381732207, 132.661743003

79208, 1.0216569111384064, 1.0931430966311486, 1.1236246179578966, 1.3009076140769134, 1 182.9283902294392, 1.354972788169064, 1.0571868150051944, 1.0077555944884016, 1.53612955 3953903, 27.212933436670724, 1.608273809125445, 37.305272888905776, 10.47561246623, 26.5 64556931244958, 8250.805195690398, 16.98595196730626, 1.5578480650208966, 1.306631436745 7972, 1.139316326800839, 1.5615633275746978, 13.07064493414851, 3.004074592255722, 240.6 806823239225, 1.1520954946328381, 1255.3802633944526, 21.496859989163646, 5.273361235560 98, 1.2180476057527334, 1.0649297644112807, 33.351260955753474, 871.2330551419003, 1.085 0439715346913, 1.0805804213827548, 1.036615657840119, 3.290736240337398, 1.3398950850120 62, 2.7582911194452886, 1.514450474506632, 1.4347450469488832, 31.63912447042936, 1.9620 623907532655, 25.975883891516684, 1.216166453812057, 8.456826514337886, 251.055773391152 14, 44.42794437207707, 835.7140118912125, 660.2001795600042, 755.9742734617693, 2.589045 738327255, 6.273347439798792, 1.0891159079365558, 7.621299990573801, 456.1495064437128, 4.108933481609017, 1.0062981212589903, 5.511741204113986]

Max vif is for variable no : 37

Iteration no 15

[66.35085436113991, 1746.7146062672582, 99.71089206809357, 2.3279374720952934, 48.716449 86007227, 249.47534652598182, 11.958164551124591, 320.65330575045675, 1.245801031698858 6, 1.0891799538722617, 7.323137337481381, 1.0586859416252266, 2.221153085543704, 21014.6 65441316818, 20944.30281515191, 1461.4319778590402, 4.071355896014691, 3.39307735641457 2, 6.555516665403206, 683.7853295086454, 1.0387465855165987, 132.696417015344, 132.66132 411351333, 1.0216453705402677, 1.092692527229627, 1.1022621888138602, 977.5178922153985, 3.419117656581771, 1.3530479334785632, 1.0571790876287803, 1.0075579826286076, 1.5353634 448529678, 25.55776821632492, 1.6076206298859699, 19588.33121451233, 4.6988668957042155, 962.4084295039324, 5.521119527434895, 1.5250591459777978, 1.306389111076366, 1.139110501 9027652, 1.3394711991888317, 5.544968921930741, 3.00265870219069, 240.6684161653598, 1.1 520195951887047, 1010.6048693061412, 20.85614440550964, 4.975729479485475, 1.19170606999 0877, 1.0646213863264373, 30.470189919631864, 8.511093871043398, 1.084434650432618, 1.08 05757373093918, 1.0365246258759626, 3.271773752157223, 1.3985632633338927, 2.75790259022 12293, 1.514317674085993, 1.4332101057935227, 31.497892728244487, 1.960936884319622, 25. 969177004420008, 1.2161599471058626, 8.332948730022705, 250.89174934161582, 44.295929368 38657, 4.275523644985996, 2.7617594715385247, 7.090433654963933, 1325.8990939707535, 3.0 629464419236547, 1.087486135332116, 5.205323555281466, 437.7216098787201, 4.108603209478 164, 1.006194989278596, 5.48831649472902]

Max vif is for variable no : 13

Iteration no 16

[66.34965234981247, 1746.4257245109932, 99.70434012327502, 2.3279289959538723, 48.712652 23236871, 249.47429321938333, 11.954589153515556, 320.6522881068839, 1.2451159829461929, 1.0873388030605062, 7.323119786606868, 1.0586774026311114, 2.2203997967474134, 109.20643 568381601, 30.209217980543038, 4.069944149390069, 3.3930607032234716, 6.555509256503007, 21.44787882437558, 1.0387257775890355, 132.69639036255757, 132.6613888476939, 1.02163688 76517647, 1.0925526203864722, 1.1021894761710866, 977.2846930377124, 3.4191157054238177, 1.353042945502976, 1.0569568496189823, 1.0075574943381054, 1.5353633588458568, 25.557224 035537185, 1.6076148883745738, 19570.505216776564, 4.697766512622495, 26.47638898231568, 5.516527333758296, 1.5236755528526127, 1.3063039276193944, 1.138906679291027, 1.33946041 06841432, 5.540585953987967, 3.002393787797348, 240.63639184769545, 1.1520194222218323, 1009.8867606399065, 20.846234494630544, 4.9753643521840685, 1.1916720024698093, 1.061574 0589331863, 30.470181330956287, 8.511060603320061, 1.0843573471907673, 1.080487847929811 2, 1.0364648983268538, 3.2634070114388742, 1.3390260578476947, 2.752488386867533, 1.5137 607041900192, 1.4327751739668322, 31.497932068866604, 1.9599521435796827, 25.96845100731 589, 1.2161511480947906, 8.33154951620421, 250.58174249204188, 44.27483640982572, 4.2741 06984587284, 2.7616206182879997, 7.090427530203092, 2.232212390009341, 6.25914264433276 1, 1.0872598365652564, 5.198034187332672, 437.41269391843275, 4.108437811415779, 1.00615 9903190288, 5.488311339807558]

Max vif is for variable no : 33

Iteration no 17

[66.34315772274273, 1746.3189538023519, 99.70082832873351, 2.327818258843756, 48.6104106 489278, 249.2728301975468, 11.95362810421759, 320.10218839520394, 1.2450899767363937, 1. 083551196502853, 7.323086598723832, 1.0586499037919503, 2.219002665070272, 109.001833306 09636, 30.139124522841804, 3.981383385647353, 3.3930101713073597, 6.502490337002233, 21. 44410789130667, 1.037652268505332, 132.68950534397592, 132.6422172739279, 1.021636776375 1104, 1.0900079558581237, 1.0996564186832292, 972.9903832230781, 3.3801411585288634, 1.3 530378522329058, 1.0569497060279374, 1.007554283585057, 1.5352173175527157, 25.556575030 75211, 1.6076008870547194, 2.4372599293000636, 26.473328832245823, 5.496256888244357, 1. 491174088253438, 1.3062695364113381, 1.1389037590925732, 1.3379586417670575, 5.082916017 424193, 3.000916140270502, 240.61609425983016, 1.1518616324162196, 1009.5311260668692, 2

0.665847904155374, 4.970925262533904, 1.1905416252572583, 1.0604394710866707, 30.4575933 592071, 8.511004680641703, 1.081856041325787, 1.080419293222878, 1.0363450672460583, 3.2 549152550772606, 1.3389068336760348, 2.7432295744845905, 1.5124641977774627, 1.432756440 7295201, 31.49267992265153, 1.9559025829096754, 25.955853499290257, 1.216015420161508, 8.324768724672195, 247.71860286363446, 41.678062770781274, 4.269470064817962, 2.76098428 7533018, 7.041803232012508, 2.198713501562878, 6.257541815126813, 1.0849338740394645, 5. 189734838655124, 420.1274064615546, 4.107336100147606, 1.0057851411091123, 5.48785983361 123]

Max vif is for variable no : 1

Iteration no 18

[4263.390658706101, 60.87529024858191, 2.278316090938073, 48.529283730927936, 247.262695 95460911, 11.946144162765364, 318.7126467186556, 1.2444558729891892, 1.0835015149225022, 7.271646290245779, 1.0585716950896387, 2.202717648632958, 108.68085242902627, 1279.02901 6389257, 3.950661757873737, 3.377554454792934, 6.2398234294170685, 650.1946128503014, 1. 037404157138327, 132.62601196758018, 132.5620711204406, 1.0214988219641623, 1.0890291109 298735, 1.0918004451688368, 972.0186935042201, 3.356740491585134, 1.3530363355676065, 1. 05691229821832, 1.007507077448131, 1.5306487957740513, 25.52983156070749, 1.607595582594 6286, 2.4356831968312034, 832.0020692581152, 5.478388208042595, 1.4909089157531554, 1.30 6199671799034, 1.1386720899654899, 1.2958024361105154, 5.082699366034001, 2.997189356094 992, 240.01036157043208, 1.1517570361050629, 996.0152184952541, 20.66516854001724, 4.960 871344429794, 1.1903332489096543, 1.060393510554361, 30.405564264460708, 8.4588725833872 01, 1.0814442920078227, 1.0803662458587748, 1.0361830529220881, 3.161714092186006, 1.398 3124868850885, 2.694270203269081, 1.5121318679808007, 1.4326777449494323, 31.25905308149 3988, 1.1453195953721313, 25.89014115380038, 1.2160046255885002, 8.319582530571493, 245. 68234902491955, 41.21869496576385, 4.252033214325837, 2.758485263950586, 7.0324132550958 69, 2.19314448725786, 6.25754064588765, 1.0845854612734853, 5.1712245826192875, 416.0555 921568141, 4.105433684876852, 1.0057831525880896, 5.485975790367248]

Max vif is for variable no : 0

Iteration no 19

[5.625326310512537, 2.277574422615041, 48.39150071662698, 247.0006305573302, 11.93504299 6644588, 318.6830661353952, 1.2444191169377559, 1.0803762514708612, 7.271608634137624, 1.057999509894599, 1.9734963957472302, 108.66497760638072, 24.1848346630639, 3.929108815 5173983, 3.3775483647026823, 6.17191814631024, 19.41833486661911, 1.0370060396018201, 13 2.62526341334964, 132.5618236947697, 1.021482698510292, 1.0889924915798495, 1.0886964997 718989, 969.8446818261688, 3.353655408891818, 1.3530312287525537, 1.056911033631205, 1.0 071220252451791, 1.5306265490114628, 25.52929608381105, 1.6074762820258346, 2.4341510185 757205, 809.2974369795396, 5.474913344544275, 1.4909022543043728, 1.3060364190903508, 1. 1373676468300602, 1.2947620825759532, 5.0814612119039975, 2.9968478983172595, 239.980958 65944194, 1.1517508471772417, 993.6995757364742, 20.643473007220667, 4.956703640459278, 1.1902995915880958, 1.0602869339388843, 30.39472311454038, 8.455514140987018, 1.08133786 84709896, 1.0803424052608759, 1.0360465427701673, 3.098347498119244, 1.3983106682779376, 2.6929663287302317, 1.5121313651367962, 1.4326410008914043, 31.246537750781425, 1.145294 7966466196, 25.890115118098105, 1.2158866527158962, 8.318940779249024, 245.644354796727 3, 41.199919357930376, 4.249704094268393, 2.7584348632793523, 6.936908275138071, 2.18709 52477818366, 6.256746724535684, 1.075836277551419, 5.162046598879537, 415.9068517829457, 4.105161031789799, 1.0054993759737452, 5.483897703504229]

Max vif is for variable no : 42

Iteration no 20

[5.611627129407225, 2.277255429201203, 47.8436844568995, 246.02320131917784, 11.86077640 1697349, 318.59535474171213, 1.2410022479973253, 1.0695374580922206, 7.270644073486629, 1.0576245048401673, 1.9685473045715758, 107.89476739351048, 1167.3840121558317, 3.919087 6582341794, 3.3763068993007277, 6.115310875152753, 618.8422948105656, 4.60578566737211, 132.6223287325534, 132.5555178196256, 1.0214820601599428, 5.023093457668223, 1.087786251 9961372, 969.8446716054382, 3.351563789987036, 1.3044543944313087, 1.053165959778511, 1. 0067616821110674, 1.5239915953165228, 25.52220072367134, 1.606968734385196, 2.4336552927 900756, 808.185221760368, 5.418985886078235, 1.488015987463004, 1.3057337240057822, 1.10 65745457922553, 1.271289348161954, 5.080369281913142, 2.993872480299661, 224.93798571177 217, 1.1514722754828182, 15.683381884371931, 4.849775488335277, 1.170935274135407, 1.058 508871022112, 29.22443121095867, 8.45549440662463, 1.080396971861612, 1.080082657233839 6, 1.035744001473718, 3.0954289417422194, 1.3981048334568889, 2.6877069919566483, 1.4822 488156524367, 1.430224927902393, 31.220224297599078, 1.1451024317256033, 25.883481096544 934, 1.2136118585416158, 8.209936594908376, 245.64437692089462, 41.18744230019773, 4.245 476983467182, 2.7584147693632732, 6.932163075695102, 2.184928265439487, 5.11704474509879 2, 1.075714129150089, 5.160371675271318, 415.90270128264103, 4.1042475571884784, 1.00542 72220203624, 5.483838494895911]

Max vif is for variable no : 12

Iteration no 21

[4.6073367925993685, 2.273752715164835, 47.80612700044493, 244.70686854131924, 11.851873 600798886, 316.97531468205784, 1.2403143078620766, 1.6720400815942416, 7.27031056863280 2, 1.056994557193586, 1.9206806520422215, 88.38774644948289, 3.861201631174597, 3.372395 0599299943, 5.835841995153146, 17.554813021838317, 1.0365798330541285, 132.6217140535664 5, 132.55528839043862, 1.0210900821110078, 1.0884824676168476, 1.083323703767467, 969.84 37221748172, 3.329613435098038, 1.3042239887373712, 1.053115687809663, 1.006525681319520 5, 1.5239820020151618, 25.479115064840986, 1.6065484637692238, 2.4328649183111484, 19.07 2917686912575, 5.41878303936455, 1.4878826874927735, 1.3054868764131655, 1.1065541798160 25, 1.2664396961472515, 5.07695726685944, 2.988325941049879, 224.47078469918932, 1.15147 20844843649, 15.677603135992735, 4.845813339292774, 1.170934371000924, 1.058457545641557 3, 29.122393642151835, 8.451394572942142, 1.0802164931207274, 1.0794231847783107, 1.0330 473384365122, 2.9984347816760635, 1.3384859105688238, 2.687670166405341, 1.4822430806715 636, 1.4301096735069558, 31.13636774324512, 1.1450263567424848, 25.869121815920018, 1.21 34169248038467, 8.2048058735095, 245.3835486516064, 41.15377871609931, 4.24203539871497 9, 2.758311202740594, 6.926435938644666, 2.182347423861119, 5.115449092448594, 1.0754500 695655302, 5.155699328309393, 415.40178762057195, 4.104246467657069, 1.0053625711993228, 5.483815411410104]

Max vif is for variable no : 22

Iteration no 22

[4.602471814316678, 2.2735298984881234, 47.600114627398604, 244.2422055245391, 11.828005 093059963, 316.898070067725, 1.2402367396493272, 1.6712246893626705, 7.268745862840594, 1.0568504622293207, 1.9186078101209065, 88.36752412813625, 3.3830207051545647, 3.3722781 175315664, 5.789145614951657, 16.445108146306154, 1.0347824511206738, 132.5705030428274 5, 132.55513889143185, 1.0209763255736002, 5.0215819845132215, 1.0832723564974789, 3.315 8354996380908, 1.3042024459524348, 1.053062796255451, 1.0065239054253319, 1.523958216430 5023, 25.47890332599904, 1.6063990400069126, 2.4328578210367966, 18.340929733621312, 5.4 05246995692816, 1.4876413236322412, 1.3054186097481364, 1.1061692031008106, 1.2663474596 049298, 5.056494391165809, 2.98831458216716, 224.0351030983636, 1.1514490629030547, 15.6 72571826454131, 4.842375196258827, 1.1708899705983813, 1.0584106558595598, 29.1188512891 57735, 8.440505859223503, 1.0802164917105952, 1.07941623434694, 1.0326493209661987, 2.99 09034531255956, 1.338422917930826, 2.6875377510526643, 1.4818356815256597, 1.42987424458 3056, 31.118015008310998, 1.1445111468280667, 25.86910935904626, 1.2134157633042946, 8.2 03501514086884, 245.3739123773641, 41.15112377236272, 4.23626717427274, 2.75814171506508 5, 6.794713575771363, 2.165731346169098, 5.1146967193041535, 1.0752818029898106, 5.15569 93897695405, 415.40139660754284, 4.1042395538084735, 1.0053618529937147, 5.4833978987832

Max vif is for variable no : 68 Iteration no 23

[376.3260055548305, 2.273529307302779, 47.02189187343344, 244.1375943838455, 11.74069559 7888568, 316.05125965444694, 1.2401994770143343, 1.0673968245334338, 7.2656522981412, 1. 033008673550566, 1.136161641651308, 88.2489794381257, 3.3764907116549567, 3.369709214170 665, 5.7728264069476305, 16.441706342478874, 1.0347775689281293, 132.57048906444942, 13 2.5550980525664, 1.0209151004697201, 5.020751362321105, 1.0765666735835113, 3.2805144367 96686, 1.299234798908615, 1.05177388398743, 1.0064980015007834, 1.522302657014947, 25.10 203325816277, 1.463839341314538, 2.410843716812431, 18.299494118882915, 5.3995961879019 2, 1.4828182276566624, 1.3054100001111935, 1.1060868283265828, 1.2641340232913296, 5.055 791269739354, 2.977146571374702, 1.4836181913141948, 1.1496580080646486, 3.1772338468156 02, 2.6960555454470043, 1.170127756470496, 1.052811990248258, 28.862098831964634, 8.4395 142125928, 1.080151837202097, 1.069297375593684, 1.0231270383537474, 2.9800580460543293, 1.3376205754915078, 2.686659898276429, 1.4814760979326478, 1.4288036506965605, 31.116189 635811295, 1.1443535811023293, 25.866234047479846, 1.2132174885218134, 8.20341258963324 2, 6.2858878958432465, 3.153061110079778, 4.230277052128129, 2.7580245821659424, 6.79257 1537376457, 2.1519631347750314, 5.040637814248862, 1.072006829282792, 4.526360148583081, 1.0033848691335083, 1841.522864286476, 2.8744814404723527]

Max vif is for variable no : 69

Iteration no 24

[4.601066028698241, 2939.0522604534103, 272764.2827049546, 243.3815147035126, 11.6397000 78742061, 316.0369378895622, 1.240199295709237, 1.668795332763843, 5478.2152482267975, 1.033008177297762, 1.1354661523510619, 2.6390673925254498, 1142.7249029501152, 3.3692072 483906563, 93.7767559304802, 524.1190872932617, 4.596210867081719, 132.57046112380752, 1 32.55510522825742, 478.6318613194022, 1.0882215872602434, 1.0765663957842755, 1103.19685 92334106, 1.2992347873144947, 1.0530064907750678, 1.0051729639705247, 1.522301045725546 8, 6814.826437674142, 1.6063193471096702, 2.4108279137006443, 663.0284658583039, 5.39955

1186871383, 1.4857981022841484, 1.3034892600672507, 1.1060781706883984, 1.26406474797293 7, 5.055779155082336, 2.9761016639537963, 1.483609970949243, 1.151337055568076, 3.176821 642428739, 2.696045509942437, 1.1696138604302022, 1.058379606790774, 122.65712852459883, 863.8327420632409, 759.0505951333982, 1.0692682683771229, 1.0231150592667824, 2.97998785 5018743, 1.397053743768552, 5.791662254690663, 1.4804188133685887, 1.4277373694310942, 3 1.089702727289453, 1.1443104311725245, 25.866192548261925, 1.213383770300042, 119.597097 16797055, 3801.697780357412, 113.73352340284345, 4.230243116239867, 2.757988266402775, 6.792220426195524, 1278.1556239019258, 2.4666447353035297, 1.0720010759592713, 15164.525 629650054, 1.0030902970384417, 2.874479086100448]

Max vif is for variable no : 2

Iteration no 25

[4.390872069460477, 2920.58960722711, 222.7094642867551, 5.017904793129847, 207.53263797 090443, 1.2385410462265982, 1.6672249642800792, 5440.041027731315, 1.0568195310938324, 1.9129655751245642, 2.6382447688832835, 1138.1714135310983, 2.980049922991367, 93.772546 88589977, 523.7887212990489, 4.59614450494971, 132.51075922471412, 132.5541744061083, 47 7.89276144539383, 5.018916986698681, 1.0723665399952123, 1074.116880307898, 1.2831560017 356354, 1.0529586873538128, 0.975006337052288, 1.510148740596642, 5895.564033839488, 1.6 060590539602564, 2.145958246320272, 662.6945382421424, 5.1695901201535035, 1.45805314798 35375, 1.3030944633980568, 1.105244879579435, 1.2640574595193954, 4.526978960096855, 2.7 178091649788736, 1.4834510833390835, 1.1511846153052279, 15.598567307289677, 4.829443112 122823, 1.1606274895810078, 1.0583780918680348, 26.36078915119787, 863.3623996761082, 75 4.630049025796, 1.0730817198218432, 1.032169085014568, 2.9242463517693635, 1.39598031648 7893, 5.746434411578016, 1.4801778439047546, 1.427724005277637, 29.100819803368477, 1.14 41241221043943, 25.24118567620251, 1.2132713702075208, 109.91953079029422, 2825.91335286 4268, 100.06588221083948, 4.199250759936699, 2.7402121336182526, 6.694409767393865, 124 7.3946050278112, 2.3235077204877936, 1.0695696633135259, 2.9859498407651572, 4.102948998 172745, 5.47455050200938]

Max vif is for variable no : 26

Iteration no 26

[359.0404538195751, 2914.8501366032756, 205.11409641357145, 5.014986319652984, 190.14923 739531412, 1.7085342557370873, 1.0660053621548657, 6.844710660228776, 1.056176738077571, 1.9099742960369404, 88.21295138553351, 1137.328200035688, 2.8248122825985154, 92.5619316 4968306, 523.6997212739535, 4.595959658720732, 132.51060947069368, 132.54568777256293, 4 77.4355405308817, 5.018396969894794, 1.0581324307353295, 3.1891589662621858, 1.280835387 5971892, 1.0527846128584735, 0.9729968656521395, 1.4881666498190198, 1.6060431255611511, 2.1385144003230927, 660.0342576557206, 5.167739160913949, 1.4574630915571285, 1.30486899 34918332, 1.1047187872813353, 1.2640225212981544, 4.524437793501285, 2.6628756939712415, 223.94921091421537, 1.1510586125878768, 15.563099620118042, 4.828235758710412, 1.1554101 0732824, 1.0578793345721866, 18.843287115569975, 8.170129473149197, 1.0712786918832216, 1.0730666331152219, 1.032131357835231, 2.914773393091153, 1.3959706216326968, 2.66103579 86620516, 1.480170374959413, 1.424636185675618, 27.909525214728916, 1.144069317148271, 2 5.02601668630198, 1.213222133330785, 33.763447099441784, 2782.8925630623303, 90.57995724 10578, 4.197463497033514, 2.7402079943040127, 6.694175336137697, 2.0960331446460376, 2.2 996172716405456, 1.072149528375704, 2.9846905999155866, 4.1029472693617315, 5.2508256124 93061]

Max vif is for variable no : 1

Iteration no 27

[338.49960471465533, 204.9994797273494, 4.778114938492993, 190.13211022167786, 1.2232206 984528773, 1.0621864117828297, 5.607462788522864, 1.0561762975531892, 1.904661706048958 3, 88.15863999295404, 1135.7883775256519, 2.759843166278329, 92.09937927629295, 523.6593 297084688, 4.594792024035619, 132.50157619610107, 132.54207708075035, 477.4041632073787, 5.016609177483754, 1.0570765766637635, 3.184573273682152, 1.280186816613409, 1.051514463 5667366, 0.9727887991282747, 1.4463662291817165, 1.6051505328256848, 2.137620811346156, 659.5898452907907, 4.900636522669029, 1.456956280111917, 1.303430149746945, 1.1044135904 015215, 1.257979001694789, 4.515457888009489, 2.5699930033831864, 217.77607676276267, 1. 1510558624288152, 15.414605022421602, 4.823129861662074, 1.1552519651320394, 1.057682623 4933314, 18.743322178974676, 7.512483672086203, 1.0679096399936892, 1.0713900849469984, 1.0320402689664034, 2.843516545216141, 1.3929451855772053, 4.5939289448289085, 1.4800991 27708219, 1.4246218583478858, 26.27759706302779, 1.1440377239600437, 21.945232210282718, 1.2125630407292876, 33.39000570557271, 2767.870884342119, 2.5110262109006793, 4.19524132 9393318, 2.7394104287113445, 6.686644373258239, 2.0955194360570935, 2.2994481808884784, 1.0680046777486247, 2.962867768951712, 4.102823640998212, 5.25073515524447]

Max vif is for variable no : 56

Iteration no 28

[338.46931784423043, 160.32165494067266, 4.776348538598456, 146.0844471500931, 1.2222989

031844018, 1.0616677676621489, 5.547678298647474, 1.0559516005614569, 1.903845855969236 6, 87.65616170611682, 1114.553771448756, 2.505540968760031, 92.04559733535389, 523.22217 2243903, 4.594038216024725, 132.48584650599966, 132.5419214516041, 476.95808436541944, 5.016233649296594, 1.055566616688933, 3.0347019079447004, 1.2789853407143672, 1.05148550 00368577, 0.9581184933061588, 1.4405941349799536, 1.6051476020654123, 1.442710158206085 5, 658.220279037809, 4.833526545782011, 1.446882240111153, 1.3034222132498186, 1.1042789 69730538, 1.466321102263164, 4.388385924593122, 2.4636937643877803, 217.5245629317649, 1.1510022083608373, 15.41218297184783, 4.819899708809678, 1.1152806654151077, 1.05768167 30631156, 16.371663845739157, 7.466571183005133, 1.0633117880848124, 1.071382697176211, 1.0318182005063867, 2.743344229106422, 1.3928630856624205, 4.591386986143127, 1.48009877 19533394, 1.424196941985951, 23.744724535531773, 1.14350580484236, 21.69903826010551, 1. 2125456050364511, 33.378247937080445, 2.4910248283579373, 4.175512317300447, 2.739365135 717147, 6.676185450822046, 2.0858238928986963, 2.2830102176972025, 1.0676885396300828, 1.8871968479173247, 4.1027414500184705, 5.216574103097952]

Max vif is for variable no : 10

Iteration no 29

[328.4165823617992, 157.95332380103432, 4.775121259647156, 143.7859648108836, 1.22199673 7856198, 1.0616365714079012, 5.541415755688212, 1.0557104862049358, 1.898215329615006, 8 6.9004470750717, 2.4982449352156557, 5.611244070921026, 16.359089142166475, 4.5375693845 25504, 132.4826233489204, 132.53853472230514, 1.0173427414372147, 5.010540428222744, 1.0 553430729965783, 2.8373532710026317, 1.2788334090940396, 1.051184858747885, 0.9577155475 216537, 1.4402683030943213, 1.6045371838632192, 1.4374439653456261, 18.07148084019496, 4.728883866420751, 1.4462320215294087, 1.303075769757576, 1.1042480289387941, 1.46439307 0164039, 4.222803854781857, 2.4549967338683234, 216.63376917892853, 1.1509442297249703, 15.385535960552208, 4.793959437168132, 1.1139652752636107, 1.0575305237254826, 16.328583 528452125, 7.368121306718486, 1.0631711663690555, 1.0713235636330598, 1.031494325231898 7, 2.739393392114565, 1.3926919038088112, 4.571072721182908, 1.479301601235644, 1.424196 3800315933, 23.609334464498968, 1.1433821819175056, 21.691657299866577, 1.21252734223909 14, 33.361418966930415, 2.476290294340151, 4.136043593171529, 2.7353746243346584, 5.5154 15004888431, 2.045451691818112, 2.2813628993283177, 1.0673785908809477, 1.87570541831071 78, 4.102709814386062, 5.215686863953837]

Max vif is for variable no : 0

Iteration no 30

[155.2323644536732, 4.7604860631069394, 141.3592348478903, 1.2216691778441484, 1.0593232 49396747, 5.528472734159787, 1.0556193133381917, 1.8980857082424056, 81.7838923214519, 2.4866005498592867, 91.08948693570807, 500.9197613252187, 4.537114563659522, 132.4792419 625271, 132.5377763376055, 476.94925424332234, 5.006338577373165, 1.0540178198548464, 2. 8339086464308307, 1.278022015946004, 1.0510783589496628, 0.9563877452348855, 1.439566749 2233684, 1.601507793291198, 1.4373855273890024, 650.8907531209678, 4.5531019625956475, 1.4461981613501722, 1.3028711846789462, 1.1041704232903105, 1.2506434362526662, 4.159668 859777641, 2.4389775010572556, 214.6294354054244, 1.1509440091693477, 15.38550821104702 3, 4.788775031172034, 1.1137482466052655, 1.0574979863018827, 15.97902694680091, 7.29027 2994680412, 747.5266388627774, 1.0702742510507022, 1.031283139101325, 2.405780674142793 6, 1.3919916350882917, 2.0887119157518077, 1.4734972836120024, 1.4240196430724281, 23.44 054108967337, 1.9525720697046762, 21.68960581595725, 1.2116362372849, 33.32088082787743, 2.4492597445359054, 4.135497020354051, 2.735314610930994, 5.430477460691674, 1214.867217 3942118, 2.2726973606251155, 1.0673776139335989, 1.853478127899001, 4.102661965664564, 5.197122041071145]

Max vif is for variable no : 58

Iteration no 31

[151.84691234038374, 4.751306036204824, 138.19114684069234, 1.2213354433500663, 1.059276 4779280937, 5.522703824566512, 1.0556133949978237, 1.8980215470773834, 81.7259155188483 8, 2.480754019309727, 91.08241608315481, 489.95402042177085, 4.537082605980154, 132.4790 3155933284, 132.53738728364056, 476.93425719369657, 1.0848631465565486, 1.05355067095389 64, 2.758621970642782, 1.2779556583864378, 1.0510086093798379, 0.9559628978306931, 1.439 099701004403, 1.6014920116114328, 1.3307079619909001, 640.6714031877855, 4.4354184683782 15, 1.4461559565909952, 1.3027844550540164, 1.1040176272468107, 1.4578986096162247, 3.94 130572148501, 2.421473257213252, 214.06464923763153, 1.1509300805036566, 15.365906682788 161, 4.786235906908415, 1.1092223271748085, 1.0574968008608283, 15.97024686219733, 7.278 882490000493, 1.0627264407580959, 1.0702353854349835, 1.0310894913942958, 2.405514460946 0015, 1.3918557756767902, 2.0869525538655687, 1.4734853258866907, 1.4238641497146578, 2 3.246161463899966, 1.1428599843427443, 21.684251968690784, 1.2114597854017768, 33.320598 203907444, 2.4055740175001947, 559.11028649222, 631.8249994891019, 5.382337994806531, 4. 643634209339676, 1.0665112504302312, 1.84834627813383, 4.102660306758133, 5.197121644097 681

Max vif is for variable no : 25 Iteration no 32 [151.83672613500767, 4.745071180377487, 138.16917585522307, 1.2213112443448004, 1.059268 825347307, 5.5220694208063055, 1.0554616720976968, 1.8979548785196538, 81.7223684479990 7, 2.479888059819891, 74.38389412894499, 178.03785896786334, 4.53688376125086, 132.47686 438393873, 132.53277753585962, 476.8744874268916, 1.0791544850780834, 1.053390332001314, 2.7235804223121773, 1.277861565511407, 1.0510079918765094, 0.9559443428171693, 1.4390811 349828105, 1.6009154675127861, 1.330542909647196, 4.428817899426784, 1.4458502173136467, 1.3025624918108953, 1.1033551153574574, 1.4541799072105168, 3.941073025138057, 2.4210108 948836306, 213.86878578157794, 1.1509203885621633, 15.361210145205076, 4.78381666046470 2, 1.1092010590502768, 1.0573641876248956, 15.96109188713119, 7.274165498783519, 1.06271 55382793356, 1.0700861278077087, 1.0254997305061466, 2.404800333044451, 1.39183242393634 15, 4.444923598669975, 1.4728024807023659, 1.4238234140865917, 23.24614853601183, 1.1425 344947800646, 21.682785564206014, 1.2112616433386654, 33.32042930586259, 2.4053389126312 62, 558.0520430613532, 631.239792852559, 5.357073544225873, 4.643584203511524, 1.0655955 886228876, 1.8480272096490464, 4.102569235778861, 5.196588876544596] Max vif is for variable no : 55 Iteration no 33 [151.7004696562861, 4.7047536519307265, 138.1141624891725, 1.2209339056524438, 1.0592227 728906356, 4154.629501102576, 1.0554495943104836, 1.8955668236824395, 2.443901490719526 7, 2.4313138740034472, 74.32704771851735, 178.00949635271087, 4.536813579379635, 132.475 05454450283, 132.52992315929743, 476.8717372140838, 1.0787948453924896, 1.05317799341617 05, 2.7194436459390428, 1.2773813422973477, 1.050947161807517, 0.9557153613646017, 1.427 7813126797507, 1.6005179361215247, 1.6495697197651598, 4.428553946054909, 1.445817226939 9865, 1.3006336322081231, 1.103334061889532, 1.2472925589479003, 3.939349322326562, 2.40 1907988732363, 213.86688283515548, 1.1509077992713388, 3.1191604079215662, 2.66606554572 43587, 1.108680839494829, 1.0572595979132173, 3.736706139951873, 6.950884680978339, 747. 3057825462251, 1.0700227949670382, 1.0254992887554293, 2.403884956792727, 1.391642889581 367, 2.0563836952884342, 1.4719060556138464, 1.4228983292594273, 23.24148005493073, 1.14 22013849843875, 21.66290492558949, 1.2112060833536582, 33.19512637848298, 2.403907609226 0685, 1.5770497701219997, 5.07278363884151, 2.2699544764327477, 1.0629911116390125, 1.84 64282397777223, 1.0029954237179421, 5.102183477037892] Max vif is for variable no : 5 Iteration no 34 [144.475773168685, 4.628092236677825, 131.6513123783874, 1.2196434993383911, 1.059160035 458322, 1.0553965679833786, 1.8945822308958074, 81.67970359535207, 2.400392179818706, 7 4.15672826792085, 177.98763703407397, 4.5362092670046215, 132.46414524988282, 132.528854 07265862, 476.7402641860449, 1.0782630313431103, 1.0529720769212723, 2.717539971711098, 1.2775651958225838, 1.0507779881494934, 0.9536498249991537, 1.2966857881246512, 1.600411 0147309816, 1.6477220617140589, 4.419393719851699, 1.4444562044925222, 1.300598104298727 6, 1.103249491520713, 1.2466029600698547, 3.9336263205593287, 2.3027962575878176, 210.31 813778214078, 1.1509077251198663, 15.353707564162551, 4.783517581004091, 1.1086741992010 563, 1.0571676308532576, 15.809229060004913, 2.9104932119337255, 746.4212392825114, 1.06 97240665760515, 1.0254786896230463, 2.3990875377667487, 1.3913836009926832, 2.0495549932 049055, 1.4715417242444766, 1.4228889626521402, 22.898429370008024, 1.1410475370648445, 21.658268172993523, 1.211196410761175, 32.490058660972736, 2.4003970731788655, 298.17090 2733994, 5.0715117587917895, 2.2699172407803823, 1.064221972915382, 1.8451130488808014, 4.102559078511201, 4.937284392095374] Max vif is for variable no : 39 Iteration no 35 [142.1224231805407, 4.6115042603108956, 129.4724363982263, 1.2196433953477275, 1.0591026 407929682, 1.0553959592324837, 1.894428669516326, 81.67219030899699, 2.3663229485181403, 74.15670168503331, 177.98084939020976, 4.536207111802764, 132.45929756058516, 132.527223 9199859, 474.3617761028861, 4.973422492873519, 1.0529720347492364, 2.717209339797263, 1. 2775642310913156, 1.0507645334817133, 0.9526081824829112, 1.296461921496262, 1.600276045 1583304, 1.3287225845827282, 4.417527795856352, 1.3671743040046132, 1.3022605360194015, 1.1032365433540956, 1.453195234806991, 3.9320266305321314, 2.2938325133486024, 210.26196 536974436, 1.1509029845079564, 15.329531383042305, 4.783361728869135, 1.109151899696855 7, 1.0570370256761197, 15.806871382368644, 2.9103242653214787, 1.0693413624958517, 1.025 47851284955, 2.3983955733956743, 1.391326801835978, 2.04923573067744, 1.471399115030444 2, 1.4228654781685692, 22.69026743892622, 1.1410310031895694, 21.645362437140786, 1.2111 833719526246, 32.48960364621818, 2.398091480724748, 298.0872580012233, 5.07017129054461

2, 4.6386107476864735, 1.0634815372096575, 1.8427029136670938, 4.1025408497783316, 4.936

Max vif is for variable no : 14

577332365134]

Group_5_Module_4_Assignment_2 Iteration no 36 [141.08553789997785, 4.611204797784756, 128.46050841322958, 1.2195380268908855, 1.059083 5222598263, 1.055373440106775, 1.8942554116574302, 81.55413340201608, 2.361767845172028 3, 74.14905271433172, 177.97869312933307, 4.536205509595707, 132.4504487801231, 132.5257 5869002283, 4.972906213047227, 1.0529702515582486, 2.7172012053183434, 1.277547000246371 8, 1.0507634153093033, 0.9523481174618491, 1.2963865330478255, 1.6002744569588208, 1.328 7183946269572, 4.417506435984275, 1.3670316368045623, 1.3022588137036217, 1.103124094454 7744, 1.4531653612243154, 3.93201157688317, 2.291259436348037, 210.26159864210632, 1.150 9023431841925, 15.32911179541859, 4.782695060366597, 1.1091453610406912, 1.0570370255664 794, 15.806785530849307, 2.9103225790770937, 1.0693395818025933, 1.025474620386796, 2.39 8362475990057, 1.3913247631555477, 4.4174623371448805, 1.4713976715442254, 1.42279658069 40253, 22.63360587474992, 1.1409501234845627, 21.641658532845906, 1.211183064482529, 32. 4852401432528, 2.3978053183633796, 298.06919510838037, 5.070134980960648, 4.638489658953 331, 1.063469751296863, 1.8417691996668204, 4.102537947703456, 4.936191625409167] Max vif is for variable no : 51 Iteration no 37 [140.40527329032633, 4.6051373473589825, 127.85829402673865, 1.2191012130751917, 1.05846 61117477572, 1.0548550221765558, 1.8940729361999045, 81.3055812840762, 2.36176655785564 8, 74.099998375445, 177.1635648978775, 4.535281776502886, 132.4504342128022, 132.5252242 0215053, 4.904529426478765, 1.0528361768390775, 2.708803059134273, 1.2775401263780675, 1.0507605191746423, 0.9523432009629057, 1.2957284615125606, 1.6002101806193665, 1.328273 8752893275, 4.410161778908017, 1.3667531021416528, 1.3020339543751513, 1.102552197147643 5, 1.4524798229544023, 3.9297204220186224, 2.2837339689181833, 210.2415243103239, 1.1508 776519500055, 15.318738426426878, 3.9802272305922086, 1.1090608898876346, 1.057036959393 647, 15.664849496007571, 2.900839069113238, 1.0693309647987614, 1.0253988678283277, 2.39 81847358472286, 1.3913020692007736, 4.238161706142234, 1.4703101684433986, 1.42278647807 09383, 22.63321644998841, 1.1409191307814623, 21.585207593916966, 1.211182408890382, 32. 27162922244017, 2.3976825946449716, 5.052232397925019, 4.6383682789116305, 1.06346795852 79605, 1.839980974523494, 4.1024731321428565, 4.907620720011942] Max vif is for variable no : 30 Iteration no 38 [140.11939995004977, 4.601106433442636, 127.55618183417481, 1.21146015274957, 1.05516617 16947487, 1.0548195861500644, 1.8939479280409544, 81.29620488894899, 2.3523006696934026, 71.14271250246745, 177.07087892885355, 4.534832153884109, 132.45042094434734, 132.524420 25275423, 4.903903884018181, 1.0526611394446854, 2.6977684311547874, 1.265726532871995, 1.0485016876810633, 0.9521767481599955, 1.2932559902707872, 1.5377444698290523, 1.328273 0221351673, 4.403323867922017, 1.3655985975104614, 1.3010156381288744, 1.097705124036208 6, 1.4507611036793788, 3.927829521123713, 2.2837223299509444, 1.1469416009516633, 14.994 229110821493, 3.96292646957852, 1.1053104480982365, 1.054873780929517, 15.47713948792781 7, 2.8811827519316418, 1.0688802310671386, 1.0253327905105936, 2.3871437252815877, 1.374 5214329397166, 1.9646911527812565, 1.4700948597180166, 1.4217689869210426, 22.6282169892 77195, 1.140734608613394, 21.58012964170364, 1.1993485278749376, 32.10887321181718, 2.39 39760984849934, 5.049792130472859, 4.553984762246713, 1.0629038937929458, 1.834905032660 8873, 4.102439250731136, 4.904253512885846] Max vif is for variable no : 10 Iteration no 39 [139.59260328944234, 4.578676354196356, 127.15500213566654, 1.2113392831144545, 1.054884 2106787213, 1.0496809479221423, 1.8869391411618937, 65.48952230173883, 2.35084744961266 1, 33.00166223263941, 4.534539384689382, 132.4353890758351, 132.49521488264685, 4.882017 110049711, 1.0515031384919225, 2.6938101489596544, 1.265703881793827, 1.048478616381526 3, 0.9521650165942495, 1.2931371465709744, 1.5376732729503997, 1.3272105093116948, 4.400 955693389555, 1.364460310897829, 1.3010151516814534, 1.097041142715842, 1.44568954803940 99, 3.9277193744267436, 2.277533022172979, 1.1468929630488385, 14.987150805908279, 3.961 7171357668184, 1.1053061940014823, 1.05472065825347, 15.46876145016925, 2.87622132497457 67, 1.0677319238541172, 1.0202341229527159, 2.38408640890488, 1.374511908832366, 4.22930 4865124899, 1.469779029966537, 1.4200839421014502, 22.59621246123255, 1.140710526879499 3, 21.577972117054173, 1.1993366301493829, 32.10854662929785, 2.393669844902885, 5.04223 7944656984, 4.552737380089638, 1.0608709709098976, 1.8200600160951408, 4.10243818893294 7, 4.903969892525972] Max vif is for variable no : 0

Iteration no 40

[4.560747289284023, 2.360186594097417, 1.2094076421285975, 1.0547783343135833, 1.0258570 005381349, 1.1165433649026884, 1.9582883222537713, 2.026570509419154, 2.017164400314814, 1.020883482428637, 132.37689962861157, 132.49403718656808, 1.0570758970992102, 1.0512537 441685235, 2.68759282741389, 1.2598839854156503, 1.0471672868815838, 0.8747707010728596,

1.2922470807736193, 1.4012771177264804, 1.3271725400016559, 4.3948232858828025, 1.353161 3582784892, 1.2986273545212061, 1.0969129187495226, 1.2401191648725158, 3.86755839359178 26, 1.5847285882338016, 1.1452198018249504, 3.042635192051277, 2.2075081044015854, 1.104 4041255260975, 1.0490743896994341, 3.6299561895039325, 2.872240899638815, 1.057998725771 7678, 1.0111094461850751, 2.2697215795128973, 1.315773539090851, 1.8771962081708904, 1.4 696441366452682, 1.4200804738790276, 18.85914915624768, 1.1402706459181005, 21.339142893 285025, 1.1990834664378094, 2.2000168874289465, 2.1944020926665186, 5.030591357605938, 2.183869826597191, 1.057706776779289, 1.49777085930682, 1.0029464947811595, 2.5671679070 030398]

Max vif is for variable no : 11

Iteration no 41

[4.560692255287168, 2.360175886704157, 1.2086742138300173, 1.0546845764504607, 1.0258553 114194229, 1.1165420858671902, 1.9581425099600462, 2.0265552712325845, 2.017123784889329 3, 1.020880152483665, 0.9239684698310372, 1.0569828718949275, 1.0512536677996096, 2.6875 629587484156, 1.2598580243649435, 1.0471627581143603, 0.8747690322576891, 1.292241051550 8518, 1.4012702037962639, 1.327159389775372, 4.3943009136355, 1.3531613525871042, 1.2906 433260473016, 1.0969115618278884, 1.2399832099649584, 3.8674410399825314, 1.584671100106 674, 1.1452191209324927, 3.0424916991053594, 2.207328346111428, 1.1044041140456968, 1.04 90712568399436, 3.62962782574712, 2.8720806217166692, 1.0579965311923805, 1.011101874021 881, 2.269614829749359, 1.3141948055459185, 1.877168142406708, 1.4696195172515287, 1.420 0503721963131, 18.859148977957403, 1.1401726577043811, 21.339126185650954, 1.19908346618 48118, 2.199998659218971, 2.1943852733906803, 5.029103306447686, 2.183665989103417, 1.05 77053572130306, 1.49775548628343, 1.0029462749946871, 2.567164986423855]

Max vif is for variable no : 43

Iteration no 42

[4.143177519600006, 1.4479020638877995, 1.208619248625275, 1.0545618632576084, 1.0258377 78081335, 1.1164173531547015, 1.9578800404261392, 1.3752768589212034, 2.017111604249464 6, 1.0208527562046625, 0.9220436864381416, 1.0563168009864023, 1.051253021492836, 2.6874 05858404476, 1.2598171264955709, 1.0463412928034967, 0.8743371443780552, 1.2909701899351 207, 1.4003796595450428, 1.3271458842852464, 4.3930947677999415, 1.3418420886778837, 1.2 906355083255658, 1.096730327864608, 1.2399539698830877, 3.867233852577633, 1.34476723734 21442, 1.1451383392960144, 3.0424901287736335, 2.207013556832733, 1.1043999987065922, 1. 049052996577832, 3.628581855158984, 2.8671365308971506, 1.0576334985188767, 1.0111007579 998594, 2.2692462565898865, 1.3141845195072142, 1.8771677058076492, 1.469619513024143, 1.4200252510248765, 14423.086837714542, 1.1399162525353135, 1.1988594602803568, 2.197634 5797221093, 2.1939865226724513, 5.027536017982444, 2.1835247534719593, 1.057570137517336 3, 1.4962275105035217, 1.002945076862517, 2.567078833789251]

Max vif is for variable no : 41

Iteration no 43

[1.1470416520817353, 0.8818228288843423, 1.2085485739188282, 1.0545339713116622, 1.04951 40109082357, 1.88405999509054, 65.40636483904707, 1.3755279387323007, 32.75232683253191 4, 1.0208501296789076, 0.9192977741073698, 1.0562176830049543, 1.0512113230108893, 2.684 8992713066697, 1.256907083547614, 1.041978880122135, 0.8642609042352009, 1.2888240918542 437, 1.5326658480288124, 1.3264590766112556, 4.392677459246204, 1.1944202412524798, 1.29 22882900936554, 1.0967300835869982, 1.444999373534515, 3.8570438268233467, 1.20548620963 54844, 1.1463547658854127, 14.98247920178043, 3.9547069276647706, 1.103533949183173, 1.0 5458443262095, 15.09283234431665, 2.862776672594182, 1.0646357637976065, 1.0202195476093 645, 2.2549871250372937, 1.3721584760158185, 4.003458213670861, 1.4689319346263252, 1.42 0024192488655, 1.1399128443998996, 1.1986058504933599, 31.842335154936254, 2.16521743757 9002, 5.022702366770009, 4.388413353593363, 1.057181760485908, 1.4503423255471863, 4.102 2831659953285, 4.790025980178837]

Max vif is for variable no : 6

Iteration no 44

[1.1468768878622662, 0.8815779693474786, 1.202603034159136, 1.0516847303565569, 1.047879 5387703677, 1.8839317710289856, 1.3754626173376434, 23.675604977071544, 4.53413357314121 6, 0.9192797488392095, 1.0550840875143477, 1.051178888741455, 2.679793193799081, 1.25690 70830512976, 1.0419771487586404, 0.8640175402345628, 1.2883837269935468, 1.5307037220394 715, 1.3257034160893029, 4.382834738274568, 1.1943584570985937, 1.2921667544592716, 1.09 05588791752023, 1.442859636715871, 3.854631712418256, 1.2054796595947481, 1.146196959886 1548, 14.940119239360286, 3.9442976992287204, 1.1033341045363014, 1.0544630554712264, 1 4.823915430077134, 2.862776081373064, 1.0632992068014975, 1.0201381636527143, 2.13468638 5768237, 1.3721204932058, 1.8527786615606778, 1.4687989699790276, 1.419664356047946, 1.1 386615675276917, 1.1984157650026799, 31.746745423597293, 2.157534814589375, 5.0156426332 66676, 4.3882288370794695, 1.057051599937959, 1.4439306370430585, 4.101549257900018, 4.7 88827285972028]

Max vif is for variable no : 42 Iteration no 45 [1.1455954189435353, 0.8802477746510878, 1.2017764996635094, 1.0514811553203645, 1.04348 34957054306, 1.8838448909389396, 1.3467126756844925, 23.665987024031327, 4.5339453786015 55, 0.9191516531682057, 1.0549231650989161, 1.0505912236550132, 2.6742454814263334, 1.25 68619461853385, 1.0406218126419136, 0.8639568638828956, 1.2572279667533703, 1.5304173736 69316, 1.3184152346366407, 4.314841323870838, 1.1943323050924795, 1.2921088084306966, 1. 087666508555954, 1.423511429810457, 3.7555296193859316, 1.2037451966166741, 1.1459882966 594677, 14.581609811584745, 3.9210602622276167, 1.1019149976600784, 1.0544380461144305, 10.78566839012119, 2.8586834954904226, 1.0624756244048532, 1.0182458270440238, 2.0879822 35288998, 1.371816264061848, 1.8330343487744731, 1.4666743486684493, 1.4195617556744773, 1.1384893259924618, 1.1984154844180683, 2.063557942976764, 5.0134714610754445, 4.3858450 94301268, 1.057039065234694, 1.4270641647378293, 4.101543589229231, 4.1029029267222] Max vif is for variable no : 7 Iteration no 46 [1.1404561140204978, 0.871050254730208, 1.1990946416446107, 1.050368348799786, 1.0433849 501419667, 1.8659166926054012, 1.3465328439553812, 4.532295380605765, 0.919020139903613 2, 1.04749813747718, 1.0505853293743046, 895.5410679930741, 1.2558142944685728, 1.040312 3179220883, 0.860843857167957, 1.2572063129758129, 1.5268604946274797, 1.633037845576110 5, 4.2653809320747, 1.193579081739733, 1.2919066776170447, 1.0876558684363076, 1.2204021 086785894, 3.755456039246217, 1.2031509613541895, 1.1446993484727146, 14.57986132254617, 3.8915014727806105, 1.1013527707309658, 1.054310673031354, 10.773316125913466, 2.8586574 965467246, 1.0599387306366526, 1.0180516160226374, 1.9638006675463187, 1.37098911352558 4, 1.824705011049574, 1.461172569051026, 1.41945192162326, 1.137965148336148, 1.19815849 31248805, 2.063183107317838, 509.1057489770541, 4.370302073269095, 1.056720975120328, 1. 4262378600375327, 4.101530566695989, 4.09496944422433] Max vif is for variable no : 11 Iteration no 47 [1.140182694332601, 0.8703059990821341, 1.1988196704316, 1.0501774889337223, 1.042273347 6178148, 1.8649637905127978, 1.3448923529747092, 4.529812942190834, 0.916310293235503, 1.0474801424791345, 1.0491410445920308, 1.2557940686028086, 1.040300293271724, 0.859740 1947353208, 1.2524371111934312, 1.52615773960554, 1.3158895405472075, 12.40605990287973 8, 1.1934262672055609, 1.2912095068476293, 1.0873934360078803, 1.422554892803882, 7.4537 82999134831, 1.20310668490865, 1.1444886983778833, 14.569401334162443, 3.879151676937883 3, 1.0966661541872404, 1.0542012842061266, 10.76428311108788, 276.42631020309824, 1.0598 793652331127, 1.017886397104867, 1.9459269208559975, 1.3708327568870873, 3.9301395741998 877, 1.461033999644815, 1.4193683188645967, 1.1379644072506754, 1.1978868452783311, 2.05 8724245537881, 262.1677470424688, 4.368099094719859, 1.056661065756184, 1.34873070148055 67, 4.101328111576599, 4.086417002125572] Max vif is for variable no : 30 Iteration no 48 [1.1398263434269493, 0.8702026790322492, 1.1988178326616676, 1.0501289124734552, 1.04224 454852501, 1.8649355223961919, 1.3315858413938906, 4.529170248600158, 0.915631706271281 4, 1.0473379937126968, 1.0487896298127668, 1.2556316439181945, 1.0402859210450115, 0.859 6403985486764, 1.1840644005914431, 1.5261189794706262, 1.3158562463920678, 12.4010879065 80965, 1.1933336737794649, 1.291179265441741, 1.0872183152491408, 1.4222690299975047, 7. 448422591027696, 1.2022827042210185, 1.1444064206863016, 14.550509261190818, 3.866817508 804468, 1.0964475267424407, 1.0541848279894508, 10.691168989607489, 1.0595189159395477, 1.0166897184777803, 1.9437186934439143, 1.369561956653677, 3.927928343861989, 1.4603384 26707351, 1.4193601260141133, 1.1377216961900949, 1.1977691013283602, 2.057302955127574 3, 138.9566159849309, 4.3657543909940495, 1.0557104489978535, 1.3484782581216739, 4.1013 20930706265, 3.7059496582832403] Max vif is for variable no : 40 Iteration no 49 [1.139635943622052, 0.8701716863355428, 1.1983506901312617, 1.0471433608831844, 1.042244 546671308, 1.8592105245869246, 1.3314226641191744, 4.519251536051953, 0.914816107154604 8, 1.0408217618284887, 1.0486387674296445, 1.2547539058237493, 1.0402575047658225, 0.859 6317447803349, 1.1824243822007012, 1.524296855058618, 1.6312348456306511, 12.40103968336 2852, 1.1933329585796517, 1.2911738323785444, 1.0867963845530717, 1.4182682330487626, 7. 445095529509006, 1.202177602133285, 1.144265239456413, 14.492664044673903, 3.64604699522 45234, 1.0961023040494497, 1.054124748479571, 10.524844467489956, 1.0588852706967933, 1. 0151160986841852, 1.785745003160463, 1.3694693722049127, 3.9136410669602126, 1.455892969

4417413, 1.4191806416710429, 1.1377016614565758, 1.1977680405750484, 2.055730727087383, 4.354455148190653, 1.0554133040623057, 1.3481403573368667, 4.1011311305265385, 3.699929

42865659251

Max vif is for variable no : 25

Iteration no 50

[1.1363209694977277, 0.8678139044557892, 1.1979572732217978, 1.0406353706782496, 1.04147 22204657114, 1.8532329631591835, 1.329397624979743, 4.5134307154712605, 0.91480890060497 08, 1.0408216729864375, 1.0482580866701883, 1.2392803016927216, 1.0351384419867329, 0.85 95130588830354, 1.180685953120282, 1.5225439508561855, 1.631172561135836, 3.900136698897 3363, 1.1872725783619105, 1.2822892678714226, 1.078495319235163, 1.3910736761743796, 7.4 449948113864215, 1.2013768142761998, 1.1424091812322046, 2.6174436992250785, 1.095845952 1282007, 1.0531393747095508, 9.572337420532211, 1.0523401602599511, 1.0147866325467934, 1.7780324162696965, 1.3640031229435157, 3.837897717377776, 1.441365786789446, 1.3368609 402496576, 1.1364536888476262, 1.197259135167034, 2.0556955758819635, 3.989019331738341, 1.0548674755624834, 1.346804992679108, 4.101042118121842, 3.6566525470611984]

Max vif is for variable no : 28

Iteration no 51

[1.131428559905233, 0.8480686878944798, 1.195880990675289, 1.0393475471622737, 1.0380929 585710803, 1.8527654243485023, 1.3243275548475382, 4.512618028792034, 0.914142615268702 9, 1.0398320243346806, 1.0444837746708862, 1.2131458440371274, 1.0315813677619292, 0.854 795608476406, 1.1611017481802772, 1.5145998452677065, 1.6284706616609472, 3.881546868333 0883, 1.1868590405051214, 1.281517583292478, 1.0784906012294322, 1.3639554537285954, 6.8 38535537281655, 1.2012353923935466, 1.1423348717292106, 2.6079001691709234, 1.0914826929 600652, 1.0476266863146662, 1.0522758113017905, 1.013044731136438, 1.7520967090325936, 1.3579806829131573, 3.83776820521023, 1.4069460101266398, 1.3368606387505246, 1.1343894 945744248, 1.1971029328615772, 1.9905185879079403, 3.576167419494417, 1.054641063358826 3, 1.335833539719264, 4.10102251562703, 3.0697089410515033]

Max vif is for variable no : 22

Iteration no 52

[1.131420994620163, 0.847716155391259, 1.1956947325438037, 1.0381284830372268, 1.0362783 011401788, 1.852760078511525, 1.324265023406027, 4.509253198335249, 0.9127060717865682, 1.0393014337465363, 1.034106739236278, 1.2064433460570547, 1.0314604886025585, 0.853584 9650593519, 1.1610271018357354, 1.5143553524287787, 1.621352127073525, 1.648032374084769 8, 1.186264115856146, 1.2814584691829851, 1.0783697579312659, 1.1700181463510184, 1.2005 31550724374, 1.1423332081347664, 2.603600539503169, 1.0697897668799576, 1.04757057651568 64, 1.050962915207719, 1.0129641265689058, 1.7049886296133512, 1.357096389416391, 1.7625 557470717832, 1.4022337796303532, 1.3361348382216955, 1.1332464070440098, 1.196985123367 3867, 1.7156163903879522, 1.6827606483661766, 1.054343310887894, 1.306856065999612, 4.10 1022470449611, 3.0694299717394227]

Max vif is for variable no : 7

In [23]: select features.head()

Out[23]:

| | Non-industry income and expenditure/revenue | Continuous interest rate (after tax) | Operating Expense Rate | Research and development expense rate | Interest- bearing debt interest rate | Tax rate (A) | Revenue Per Share (Yuan ¥) | Realized Sales Gross Profit Growth Rate | t G |
|---|---|---|---------------------------|---------------------------------------|--|--------------------|-------------------------------------|--|--------|
| 0 | 0.302646 | 0.780985 | 1.256969e-04 | 0.0 | 0.000725 | 0.0 | 0.017560 | 0.022102 | 0.6 |
| 1 | 0.303556 | 0.781506 | 2.897851e-04 | 0.0 | 0.000647 | 0.0 | 0.021144 | 0.022080 | 0.6 |
| 2 | 0.302035 | 0.780284 | 2.361297e-04 | 25500000.0 | 0.000790 | 0.0 | 0.005944 | 0.022760 | 0.0 |
| 3 | 0.303350 | 0.781241 | 1.078888e-04 | 0.0 | 0.000449 | 0.0 | 0.014368 | 0.022046 | 0.0 |
| 4 | 0.303475 | 0.781550 | 7.890000e+09 | 0.0 | 0.000686 | 0.0 | 0.029690 | 0.022096 | 0.6 |

5 rows × 42 columns

In [24]:

X = select features

```
In [25]:
           from sklearn.feature_selection import SelectKBest, mutual_info_classif, f_regression, c
           # define feature selection
           fs = SelectKBest(score func=f regression, k=10)
           # apply feature selection
           X_selected = fs.fit_transform(X, y)
           print(X selected.shape)
           (6819, 10)
In [26]:
           # create filter mask to identify feature names
           filter = fs.get_support()
           features = array(X.columns)
           print(features[filter])
           [' Tax rate (A)' ' Fixed Assets Turnover Frequency' ' Cash/Total Assets'
             Cash/Current Liability' ' Retained Earnings to Total Assets' Total expense/Assets' ' Equity to Long-term Liability'
            ' Current Liability to Current Assets'
           " Net Income to Stockholder's Equity" ' Equity to Liability']
In [27]:
           # create list of columns to drop from df
           columns to remove = X.columns.values[np.logical not(filter)]
           columns to remove
Out[27]: array([' Non-industry income and expenditure/revenue',
                    Continuous interest rate (after tax)', ' Operating Expense Rate',
                    Research and development expense rate',
                   ' Interest-bearing debt interest rate',
                   ' Revenue Per Share (Yuan ¥)',
                   ' Realized Sales Gross Profit Growth Rate',
                    After-tax Net Profit Growth Rate', ' Total Asset Growth Rate',
                    Net Value Growth Rate', 'Current Ratio', 'Quick Ratio', Interest Expense Ratio', 'Total debt/Total net worth',
                    Long-term fund suitability ratio (A)',
                    Contingent liabilities/Net worth', 'Total Asset Turnover',
                   ' Accounts Receivable Turnover', ' Average Collection Days',
                   ' Inventory Turnover Rate (times)', ' Revenue per person',
                   ' Allocation rate per person', ' Quick Assets/Current Liability',
                   ' Inventory/Current Liability',
                   ' Long-term Liability to Current Assets',
                  ' Total income/Total expense', ' Current Asset Turnover Rate', ' Quick Asset Turnover Rate', ' Cash Turnover Rate', ' Fixed Assets to Assets', ' Total assets to GNP price',
                    Degree of Financial Leverage (DFL)'], dtype=object)
In [28]:
           # remove features from data
           df1 = X.drop(columns=columns to remove)
           df1.info()
           <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 6819 entries, 0 to 6818
          Data columns (total 10 columns):
           #
                Column
                                                           Non-Null Count Dtype
           ---
           0
                 Tax rate (A)
                                                           6819 non-null
                                                                             float64
                 Fixed Assets Turnover Frequency
                                                                            float64
           1
                                                           6819 non-null
                                                           6819 non-null
                                                                            float64
           2
                 Cash/Total Assets
```

```
6819 non-null
                                                            float64
 3
     Cash/Current Liability
                                           6819 non-null
      Retained Earnings to Total Assets
                                                            float64
 4
 5
                                                            float64
      Total expense/Assets
                                           6819 non-null
      Equity to Long-term Liability
                                           6819 non-null
                                                            float64
      Current Liability to Current Assets 6819 non-null
                                                            float64
 7
      Net Income to Stockholder's Equity
                                                            float64
 8
                                           6819 non-null
 9
      Equity to Liability
                                           6819 non-null
                                                           float64
dtypes: float64(10)
```

Removing Anomalies with Isolation Forests

memory usage: 532.9 KB

```
In [29]:
          df2 = df1
In [30]:
          #Isolation Forest Identifies anomalies
          model=IsolationForest(n estimators=100, contamination=float(.05), random state=42)
          model.fit(df1)
         C:\Users\sjue\Anaconda3\lib\site-packages\sklearn\base.py:450: UserWarning: X does not h
         ave valid feature names, but IsolationForest was fitted with feature names
           warnings.warn(
Out[30]:
                              IsolationForest
         IsolationForest(contamination=0.05, random_state=42)
In [31]:
          df2['scores'] = model.decision function(df1)
          df1 = X.drop(columns=columns to remove)
In [32]:
          #Removing anomolies from the dataset
          df2['anomaly score'] = model.predict(df1)
          df3 = pd.concat([y, df2], axis=1)
          df4 = df3[df3['anomaly_score']!=-1]
          df5 = df4.drop(columns = ['scores', 'anomaly score'])
          df5.head(10)
Out[32].
                                                                                            Equity
```

| Out[32]: | | Bankrupt? | Tax rate (A) | Fixed Assets Turnover Frequency | Cash/Total Assets | Cash/Current Liability | Retained Earnings to Total Assets | Total expense/Assets | to Long- term Liability | |
|----------|---|-----------|-----------------|---------------------------------------|----------------------|---------------------------|--|----------------------|-------------------------|---|
| | 0 | 1 | 0.000000 | 1.165007e-04 | 0.004094 | 0.000147 | 0.903225 | 0.064856 | 0.126549 | C |
| | 1 | 1 | 0.000000 | 7.190000e+08 | 0.014948 | 0.001384 | 0.931065 | 0.025516 | 0.120916 | C |
| | 3 | 1 | 0.000000 | 9.150000e+09 | 0.018851 | 0.001011 | 0.906902 | 0.024161 | 0.120760 | C |
| | 4 | 1 | 0.000000 | 2.935211e-04 | 0.014161 | 0.000680 | 0.913850 | 0.026385 | 0.110933 | C |
| | 6 | 0 | 0.000000 | 1.452476e-04 | 0.022989 | 0.001050 | 0.911441 | 0.092802 | 0.120561 | C |

| | | Bankrupt? | Tax rate (A) | Fixed Assets Turnover Frequency | Cash/Total Assets | Cash/Current Liability | Retained Earnings to Total Assets | Total expense/Assets | Equity to Long- term Liability | |
|----------|----|-----------------------------------|-----------------|---------------------------------------|----------------------|---------------------------|--|-------------------------|--|---|
| | 7 | 0 | 0.157607 | 5.182010e-04 | 0.014616 | 0.000888 | 0.935449 | 0.025482 | 0.114508 | C |
| | 8 | 0 | 0.005871 | 2.458618e-04 | 0.145189 | 0.013190 | 0.935200 | 0.029884 | 0.110933 | C |
| | 9 | 0 | 0.392449 | 7.383418e-04 | 0.008773 | 0.000491 | 0.938290 | 0.035891 | 0.115499 | C |
| | 10 | 0 | 0.000000 | 1.747510e-04 | 0.002057 | 0.000336 | 0.942829 | 0.028802 | 0.120943 | C |
| | 11 | 0 | 0.000000 | 1.066662e-04 | 0.020542 | 0.002254 | 0.932955 | 0.056952 | 0.116459 | C |
| | 4 | | | | | | | | | • |
| In [33]: | Χ. | = df5.drop shape = df5['Ban | | | | | | | | |

Data Prep

Split the Data for Training

```
In [34]: X = df5.drop(columns=['Bankrupt?'], axis=1)
X.shape

Out[34]: (6478, 10)

In [35]: y = df5['Bankrupt?']

In [36]: # split data in to training and test data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=4)
```

Scale the Data

SMOTE for Training Data

```
In [38]: from imblearn.over_sampling import SMOTE
```

```
In [39]:
    sm = SMOTE(sampling_strategy='minority', random_state=42)
    X_train, y_train = sm.fit_resample(X_train, y_train)
```

Models

Support Vector Machine (SVM)

```
In [40]:
        # # SVM Classifier model
         svm_clf = SVC(kernel="rbf", C=1, probability=True)
        svm clf.fit(X train, y train)
        svm_model = svm_clf.fit(X_train, y_train)
In [41]:
        # kfold cross validation
        score = cross val score(svm model, X train, y train, cv=10, verbose=3)
        score.mean()
        [Parallel(n jobs=1)]: Using backend SequentialBackend with 1 concurrent workers.
        [CV] END ..... score: (test=0.902) total time=
        [Parallel(n jobs=1)]: Done  1 out of  1 | elapsed:
                                                       7.4s remaining:
                                                                        0.0s
        [CV] END ..... score: (test=0.873) total time=
                                                                         7.2s
        [Parallel(n jobs=1)]: Done 2 out of 2 | elapsed:
                                                      14.7s remaining:
                                                                        0.0s
        [CV] END ..... score: (test=0.897) total time=
                                                                         8.4s
        [CV] END ..... score: (test=0.906) total time=
                                                                         7.6s
        [CV] END ..... score: (test=0.892) total time=
                                                                         7.3s
        [CV] END ..... score: (test=0.897) total time=
                                                                         7.2s
        [CV] END ..... score: (test=0.898) total time=
                                                                         6.7s
        [CV] END ..... score: (test=0.897) total time=
                                                                         6.7s
        [CV] END ..... score: (test=0.910) total time=
                                                                         6.9s
        [CV] END ..... score: (test=0.894) total time=
                                                                         6.7s
        [Parallel(n_jobs=1)]: Done 10 out of 10 | elapsed: 1.2min finished
Out[41]: 0.8966787276715028
In [42]:
        # predictions
        y pred = svm model.predict(X test)
In [43]:
        # precision and recall scores
        print("Precision: {:.2f}%".format(100 * precision_score(y_test, y_pred)))
        print("Recall: {:.2f}%".format(100 * recall score(y test, y pred)))
        Precision: 15.18%
        Recall: 82.86%
In [44]:
        # f1 score
        f1_score(y_test, y_pred)
Out[44]: 0.2566371681415929
```

Tuning SVM Model Hyperparameters

```
In [45]:
```

```
# create grid search cross validation to tunr hyperparmeters of SVC model
param_grid = {'C': [0.1, 1, 10, 100], 'gamma': [1,0.1,0.01,0.001]}
grid = GridSearchCV(SVC(probability=True, kernel='rbf'),param_grid,refit=True,verbose=2
grid.fit(X_train,y_train);
```

```
Fitting 5 folds for each of 16 candidates, totalling 80 fits
9.7s
9.7s
[CV] END ......C=0.1, gamma=1; total time=
                          9.7s
9.9s
11.4s
7.8s
[CV] END ....... C=0.1, gamma=0.1; total time=
                          9.5s
8.3s
7.5s
7.8s
9.0s
8.1s
8.2s
[CV] END ......C=0.1, gamma=0.01; total time=
                          8.2s
[CV] END .......gamma=0.01; total time=
                          8.2s
[CV] END ........ C=0.1, gamma=0.001; total time=
                         11.4s
11.3s
[CV] END .......total time=
                         11.4s
[CV] END ........ C=0.1, gamma=0.001; total time=
                         11.4s
6.0s
6.1s
[CV] END ......C=1, gamma=1; total time=
                          6.2s
CV END ........ c=1, gamma=1; total time=
                          6.2s
[CV] END ......C=1, gamma=1; total time=
                          6.6s
[CV] END .....C=1, gamma=0.1; total time=
                          5.2s
5.2s
5.1s
[CV] END .......total time=
                          5.2s
[CV] END .......total time=
                          5.3s
6.5s
[CV] END ......C=1, gamma=0.01; total time=
                          6.6s
6.6s
6.6s
6.7s
8.2s
[CV] END ......C=1, gamma=0.001; total time=
                          8.2s
[CV] END .......total time=
                          8.3s
[CV] END .......gamma=0.001; total time=
                          8.3s
[CV] END .......total time=
                          8.3s
[CV] END ......C=10, gamma=1; total time=
                          5.3s
[CV] END ......C=10, gamma=1; total time=
                          5.3s
[CV] END ......C=10, gamma=1; total time=
                          5.9s
5.4s
[CV] END ......C=10, gamma=1; total time=
                          5.3s
[CV] END ......C=10, gamma=0.1; total time=
                          4.2s
[CV] END ......C=10, gamma=0.1; total time=
                          4.4s
[CV] END ........total time=
                          4.3s
[CV] END ........; total time=
                          4.3s
4.3s
5.7s
[CV] END .....C=10, gamma=0.01; total time=
                          5.9s
[CV] END ......C=10, gamma=0.01; total time=
                          5.8s
[CV] END ......C=10, gamma=0.01; total time=
                          5.9s
5.9s
7.0s
```

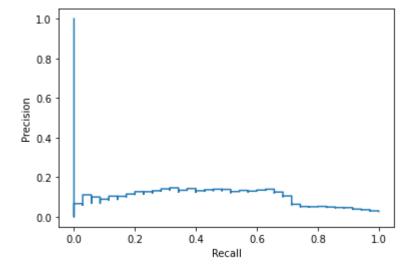
```
[CV] END ......C=10, gamma=0.001; total time=
                                    6.9s
[CV] END ......C=10, gamma=0.001; total time=
                                    7.1s
[CV] END ......C=10, gamma=0.001; total time=
                                    6.9s
[CV] END ......C=10, gamma=0.001; total time=
                                    7.0s
[CV] END .....C=100, gamma=1; total time=
                                    5.1s
[CV] END ......c=100, gamma=1; total time=
                                    5.5s
[CV] END ......C=100, gamma=1; total time=
                                    5.5s
[CV] END ......C=100, gamma=1; total time=
                                    5.6s
[CV] END ......C=100, gamma=1; total time=
                                    5.6s
5.1s
5.4s
[CV] END ......C=100, gamma=0.1; total time=
                                    5.4s
[CV] END ......C=100, gamma=0.1; total time=
                                    5.2s
5.4s
[CV] END ......C=100, gamma=0.01; total time=
                                    6.4s
[CV] END ......C=100, gamma=0.01; total time=
                                    6.3s
[CV] END ......C=100, gamma=0.01; total time=
                                    6.3s
[CV] END ......C=100, gamma=0.01; total time=
                                    6.5s
[CV] END ......C=100, gamma=0.01; total time=
                                    6.3s
[CV] END ....... c=100, gamma=0.001; total time=
                                    6.5s
6.6s
6.6s
6.6s
6.6s
```

In [46]: # print optimal values for C and gamma
print(grid.best_estimator_)

SVC(C=10, gamma=1, probability=True)

```
y_pred = grid.predict(X_test)
y_pred_proba = grid.predict_proba(X_test)
y_score = grid.decision_function(X_test)
```

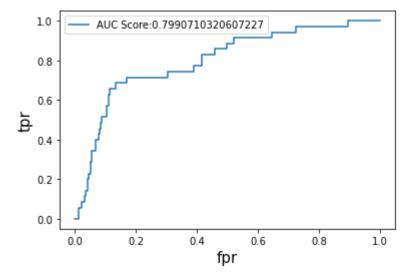
In [48]: #Precision/Recall
 prec, recall, _ = precision_recall_curve(y_test, y_score, pos_label=grid.classes_[1])
 pr_display = PrecisionRecallDisplay(precision=prec, recall=recall).plot()



```
In [49]: #Plot the ROC curve
fpr, tpr, ths = roc_curve(y_test, y_pred_proba[:,1])
```

```
auc_score = auc(fpr,tpr)
plt.plot(fpr,tpr,label="AUC Score:" + str(auc_score))
plt.xlabel('fpr',fontsize='15')
plt.ylabel('tpr',fontsize='15')
plt.legend(loc='best')
```

Out[49]: <matplotlib.legend.Legend at 0x174af22aa90>



```
In [50]:
          cnf matrix = confusion matrix(y test,y pred)
          FP = cnf matrix.sum(axis=0) - np.diag(cnf matrix)
          FN = cnf_matrix.sum(axis=1) - np.diag(cnf_matrix)
          TP = np.diag(cnf_matrix)
          TN = cnf matrix.sum() - (FP + FN + TP)
          FP = FP.astype(float)
          FN = FN.astype(float)
          TP = TP.astype(float)
          TN = TN.astype(float)
          # Sensitivity, hit rate, recall, or true positive rate
          TPR = TP/(TP+FN)
          # Specificity or true negative rate
          TNR = TN/(TN+FP)
          # Precision or positive predictive value
          PPV = TP/(TP+FP)
          # Negative predictive value
          NPV = TN/(TN+FN)
          # Fall out or false positive rate
          FPR = FP/(FP+TN)
          # False negative rate
          FNR = FN/(TP+FN)
          # False discovery rate
          FDR = FP/(TP+FP)
          # Overall accuracy
          ACC = (TP+TN)/(TP+FP+FN+TN)
          results = pd.DataFrame(columns = ['Model', 'TPR', 'FPR', 'precision', 'recall', 'accura
          #eval
          Model = 'SVM'
          TPR = [round(num, 2) for num in TPR]
```

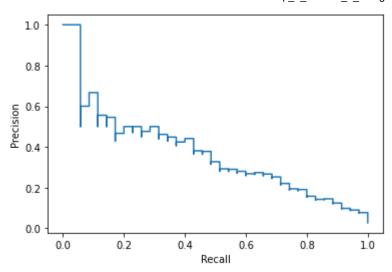
```
FPR = [round(num, 2) for num in FPR]
precision = [round(num, 2) for num in PPV]
recall = [round(num, 2) for num in TPR]
accuracy = [round(num, 2) for num in ACC]
f1_value = round(f1_score(y_pred, y_test),2)
row = [Model, TPR, FPR, precision, recall, accuracy, f1_value]
results = results.append(pd.DataFrame([row], columns=results.columns), ignore_index=Tru
results
```

```
Out[50]: Model TPR FPR precision recall accuracy f1-value

0 SVM [0.97, 0.09] [0.91, 0.03] [0.97, 0.08] [0.97, 0.09] [0.95, 0.95] 0.08
```

Logistic Regression Model

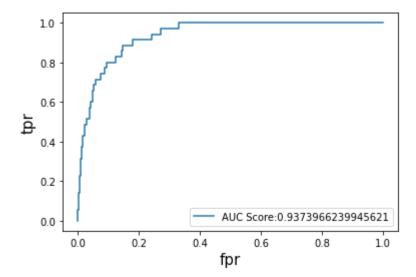
```
In [51]:
         # create logistic model
         log clf = LogisticRegression(C=1.0, penalty='12', solver='newton-cg')
         log_model = log_clf.fit(X_train, y_train)
In [52]:
         # kfold validation
          score = cross val score(log clf, X train, y train, cv=3, verbose=3)
         score.mean()
         [CV] END ..... score: (test=0.850) total time=
                                                                                0.0s
         [CV] END ..... score: (test=0.845) total time=
                                                                                0.0s
         [CV] END ..... score: (test=0.855) total time=
                                                                                0.0s
         [Parallel(n_jobs=1)]: Using backend SequentialBackend with 1 concurrent workers.
         [Parallel(n_jobs=1)]: Done    1 out of
                                              1 | elapsed:
                                                             0.0s remaining:
                                                                               0.0s
         [Parallel(n jobs=1)]: Done
                                    2 out of
                                              2 | elapsed:
                                                             0.0s remaining:
                                                                               0.0s
                                                             0.0s finished
         [Parallel(n jobs=1)]: Done
                                    3 out of
                                              3 | elapsed:
Out[52]: 0.8501087474292569
In [53]:
         y pred = log model.predict(X test)
         y pred proba = log clf.predict proba(X test)
         y score = log clf.decision function(X test)
In [54]:
         #Precision/Recall
         prec, recall, _ = precision_recall_curve(y_test, y_score, pos_label=log_clf.classes_[1]
         pr display = PrecisionRecallDisplay(precision=prec, recall=recall).plot()
```



```
In [55]: fpr, tpr, ths = roc_curve(y_test, y_pred_proba[:,1])

auc_score = auc(fpr,tpr)
plt.plot(fpr,tpr,label="AUC Score:" + str(auc_score))
plt.xlabel('fpr',fontsize='15')
plt.ylabel('tpr',fontsize='15')
plt.legend(loc='best')
```

Out[55]: <matplotlib.legend.Legend at 0x174afc71850>



```
In [56]:
    cnf_matrix = confusion_matrix(y_test,y_pred)

    FP = cnf_matrix.sum(axis=0) - np.diag(cnf_matrix)
    FN = cnf_matrix.sum(axis=1) - np.diag(cnf_matrix)
    TP = np.diag(cnf_matrix)
    TN = cnf_matrix.sum() - (FP + FN + TP)

    FP = FP.astype(float)
    FN = FN.astype(float)
    TP = TP.astype(float)
    TN = TN.astype(float)

    # Sensitivity, hit rate, recall, or true positive rate
    TPR = TP/(TP+FN)
```

```
# Specificity or true negative rate
TNR = TN/(TN+FP)
# Precision or positive predictive value
PPV = TP/(TP+FP)
# Negative predictive value
NPV = TN/(TN+FN)
# Fall out or false positive rate
FPR = FP/(FP+TN)
# False negative rate
FNR = FN/(TP+FN)
# False discovery rate
FDR = FP/(TP+FP)
# Overall accuracy
ACC = (TP+TN)/(TP+FP+FN+TN)
#eval
Model = 'Logistic Regression'
TPR = [round(num, 2) for num in TPR]
FPR = [round(num, 2) for num in FPR]
precision = [round(num, 2) for num in PPV]
recall = [round(num, 2) for num in TPR]
accuracy = [round(num, 2) for num in ACC]
f1 value = round(f1_score(y_pred, y_test),2)
row = [Model, TPR, FPR, precision, recall, accuracy, f1 value]
results = results.append(pd.DataFrame([row], columns=results.columns), ignore index=Tru
results
```

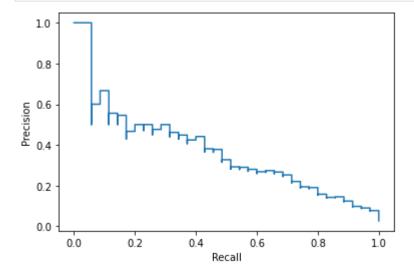
| Out[56]: | | Model | TPR | FPR | precision | recall | accuracy | f1-value |
|----------|---|---------------------|--------------|--------------|--------------|--------------|--------------|----------|
| | 0 | SVM | [0.97, 0.09] | [0.91, 0.03] | [0.97, 0.08] | [0.97, 0.09] | [0.95, 0.95] | 0.08 |
| | 1 | Logistic Regression | [0.83, 0.89] | [0.11, 0.17] | [1.0, 0.13] | [0.83, 0.89] | [0.83, 0.83] | 0.22 |

Naïve Bayes model

```
In [57]:
          # create naive bayes model
          nb clf = GaussianNB()
          nb_model = nb_clf.fit(X_train, np.ravel(y_train))
In [58]:
         # kfold cross validation
         score = cross val score(nb clf, X train, y train, cv=3, verbose=3)
         score.mean()
         [CV] END ..... score: (test=0.832) total time=
                                                                              0.0s
         [CV] END ..... score: (test=0.821) total time=
                                                                              0.0s
         [CV] END ..... score: (test=0.842) total time=
                                                                              0.0s
         [Parallel(n_jobs=1)]: Using backend SequentialBackend with 1 concurrent workers.
         [Parallel(n jobs=1)]: Done 1 out of 1 | elapsed:
                                                           0.0s remaining:
                                                                             0.0s
        [Parallel(n_jobs=1)]: Done
                                   2 out of
                                             2 | elapsed:
                                                           0.0s remaining:
                                                                             0.0s
        [Parallel(n_jobs=1)]: Done  3 out of
                                             3 | elapsed:
                                                           0.0s finished
Out[58]: 0.8315206611318201
In [59]:
         y pred = nb model.predict(X test)
         y pred proba = nb clf.predict proba(X test)
```

```
In [60]: #Precision/Recall
```

precision/Recutt
prec, recall, _ = precision_recall_curve(y_test, y_score, pos_label=nb_clf.classes_[1])
pr_display = PrecisionRecallDisplay(precision=prec, recall=recall).plot()

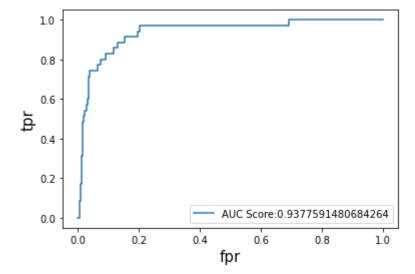


```
import matplotlib.pyplot as plt
from sklearn.metrics import roc_curve, auc, roc_auc_score

fpr, tpr, ths = roc_curve(y_test, y_pred_proba[:,1])

auc_score = auc(fpr,tpr)
plt.plot(fpr,tpr,label="AUC Score:" + str(auc_score))
plt.xlabel('fpr',fontsize='15')
plt.ylabel('tpr',fontsize='15')
plt.legend(loc='best')
```

Out[61]: <matplotlib.legend.Legend at 0x174afc45610>



```
In [62]:
    cnf_matrix = confusion_matrix(y_test,y_pred)
    FP = cnf_matrix.sum(axis=0) - np.diag(cnf_matrix)
    FN = cnf_matrix.sum(axis=1) - np.diag(cnf_matrix)
    TP = np.diag(cnf_matrix)
    TN = cnf_matrix.sum() - (FP + FN + TP)
```

```
FP = FP.astype(float)
FN = FN.astype(float)
TP = TP.astype(float)
TN = TN.astype(float)
# Sensitivity, hit rate, recall, or true positive rate
TPR = TP/(TP+FN)
# Specificity or true negative rate
TNR = TN/(TN+FP)
# Precision or positive predictive value
PPV = TP/(TP+FP)
# Negative predictive value
NPV = TN/(TN+FN)
# Fall out or false positive rate
FPR = FP/(FP+TN)
# False negative rate
FNR = FN/(TP+FN)
# False discovery rate
FDR = FP/(TP+FP)
# Overall accuracy
ACC = (TP+TN)/(TP+FP+FN+TN)
#eval
Model = 'naive bayes'
TPR = [round(num, 2) for num in TPR]
FPR = [round(num, 2) for num in FPR]
precision = [round(num, 2) for num in PPV]
recall = [round(num, 2) for num in TPR]
accuracy = [round(num, 2) for num in ACC]
f1_value = round(f1_score(y_pred, y_test),2)
row3 = [Model, TPR, FPR, precision, recall, accuracy, f1_value]
results = results.append(pd.DataFrame([row3], columns=results.columns), ignore_index=Tr
results
```

| Out[62]: | | Model | TPR | FPR | precision | recall | accuracy | f1-value |
|----------|---|---------------------|--------------|--------------|--------------|--------------|--------------|----------|
| | 0 | SVM | [0.97, 0.09] | [0.91, 0.03] | [0.97, 0.08] | [0.97, 0.09] | [0.95, 0.95] | 0.08 |
| | 1 | Logistic Regression | [0.83, 0.89] | [0.11, 0.17] | [1.0, 0.13] | [0.83, 0.89] | [0.83, 0.83] | 0.22 |
| | 2 | naive baves | [0.87, 0.86] | [0.14, 0.13] | [1.0, 0.16] | [0.87, 0.86] | [0.87, 0.87] | 0.27 |