

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
```

```
dataset = pd.read_csv('/content/drive/My Drive/kddcup99_csv.csv')
```

```
dataset.head(5)
```

```
↳
```

	duration	protocol_type	service	flag	src_bytes	dst_bytes	land	wrong_fragment
0	0	tcp	http	SF	181	5450	0	(
1	0	tcp	http	SF	239	486	0	(
2	0	tcp	http	SF	235	1337	0	(
3	0	tcp	http	SF	219	1337	0	(
4	0	tcp	http	SF	217	2032	0	(

selecting the feature from variance method

```
print(dataset.var()['src_bytes'])
print(dataset.var()['dst_bytes'])
print(dataset.var()['land'])
```

```
↳ 976576992036.6654
   1091643891.0952706
   4.453071700180602e-05
```

```
print(dataset.var()['wrong_fragment'])
print(dataset.var()['urgent'])
print(dataset.var()['hot'])
```

```
↳ 0.018172491590816044
   3.03630038804738e-05
   0.6116856844184144
```

```
print(dataset.var()['num_failed_logins'])
print(dataset.var()['logged_in'])
print(dataset.var()['lnum_compromised'])
print(dataset.var()['lroot_shell'])
print(dataset.var()['lsu_attempted'])
```

```
↳ 0.00024085837553570224
   0.12626868283082365
   3.2339838746190672
   0.0001113193556638224
   6.072508173886537e-05
```

```

print(dataset.var()['lnum_root'])
print(dataset.var()['lnum_file_creations'])
print(dataset.var()['lnum_shells'])
print(dataset.var()['lnum_access_files'])
#print(dataset.var()['lnum_outbound_cmds'])
#print(dataset.var()['is_host_login'])
print(dataset.var()['is_guest_login'])

```

```

↳ 4.051043258019034
   0.009296040477405787
   0.000121440870502456
   0.001330916397839991
   0.001384663728080717

```

```

print(dataset.var()['count'])
print(dataset.var()['srv_count'])
print(dataset.var()['serror_rate'])
print(dataset.var()['srv_serror_rate'])
print(dataset.var()['rerror_rate'])
print(dataset.var()['srv_rerror_rate'])
print(dataset.var()['same_srv_rate'])
print(dataset.var()['diff_srv_rate'])
print(dataset.var()['srv_diff_host_rate'])
print(dataset.var()['dst_host_count'])
print(dataset.var()['dst_host_srv_count'])
print(dataset.var()['dst_host_same_srv_rate'])
print(dataset.var()['dst_host_diff_srv_rate'])

```

```

↳ 45431.6891034859
   60674.89003505529
   0.1449456310521699
   0.14517386836860224
   0.0536495355281265
   0.05389232330783698
   0.15069129958002264
   0.006757755850526054
   0.020276896258574477
   4191.863370949552
   11244.525010698932
   0.1687402028545784
   0.011937575833318366

```

```

print(dataset.var()['dst_host_same_src_port_rate'])
print(dataset.var()['dst_host_srv_diff_host_rate'])
print(dataset.var()['dst_host_serror_rate'])
print(dataset.var()['dst_host_srv_serror_rate'])
print(dataset.var()['dst_host_rerror_rate'])
print(dataset.var()['dst_host_srv_rerror_rate'])

```

```

↳

```

```
0.2316583307814032
0.0017751826094751953
0.14485133584224127
0.1450000566016115
```

remove redudunant feature

```
dataset['lsu_attempted'].value_counts()
dataset.drop('lsu_attempted', axis=1, inplace=True)
dataset['urgent'].value_counts()
dataset.drop('urgent', axis=1, inplace=True)
dataset['lnum_outbound_cmds'].value_counts()
dataset.drop('lnum_outbound_cmds', axis=1, inplace=True)
dataset['is_host_login'].value_counts()
dataset.drop('is_host_login', axis=1, inplace=True)
dataset['wrong_fragment'].value_counts()
dataset.drop('wrong_fragment', axis=1, inplace=True)
dataset['hot'].value_counts()
dataset.drop('hot', axis=1, inplace=True)
dataset['num_failed_logins'].value_counts()
dataset.drop('num_failed_logins', axis=1, inplace=True)
dataset['logged_in'].value_counts()
dataset.drop('logged_in', axis=1, inplace=True)
dataset['lroot_shell'].value_counts()
dataset.drop('lroot_shell', axis=1, inplace=True)
dataset['lnum_file_creations'].value_counts()
dataset.drop('lnum_file_creations', axis=1, inplace=True)
dataset['lnum_shells'].value_counts()
dataset.drop('lnum_shells', axis=1, inplace=True)
dataset['lnum_access_files'].value_counts()
dataset.drop('lnum_access_files', axis=1, inplace=True)
dataset['is_guest_login'].value_counts()
dataset.drop('is_guest_login', axis=1, inplace=True)
dataset['serror_rate'].value_counts()
dataset.drop('serror_rate', axis=1, inplace=True)
dataset['srv_serror_rate'].value_counts()
dataset.drop('srv_serror_rate', axis=1, inplace=True)
```

```
dataset['rerror_rate'].value_counts()
dataset.drop('rerror_rate', axis=1, inplace=True)
dataset['srv_rerror_rate'].value_counts()
dataset.drop('srv_rerror_rate', axis=1, inplace=True)
dataset['same_srv_rate'].value_counts()
dataset.drop('same_srv_rate', axis=1, inplace=True)
dataset['diff_srv_rate'].value_counts()
dataset.drop('diff_srv_rate', axis=1, inplace=True)
dataset['srv_diff_host_rate'].value_counts()
dataset.drop('srv_diff_host_rate', axis=1, inplace=True)
dataset['dst_host_same_srv_rate'].value_counts()
dataset.drop('dst_host_same_srv_rate', axis=1, inplace=True)
dataset['dst_host_diff_srv_rate'].value_counts()
dataset.drop('dst_host_diff_srv_rate', axis=1, inplace=True)
dataset['dst_host_same_src_port_rate'].value_counts()
```

```

dataset['dst_host_same_src_port_rate'].value_counts()
dataset.drop('dst_host_same_src_port_rate', axis=1, inplace=True)
dataset['dst_host_srv_diff_host_rate'].value_counts()
dataset.drop('dst_host_srv_diff_host_rate', axis=1, inplace=True)
dataset['dst_host_serror_rate'].value_counts()
dataset.drop('dst_host_serror_rate', axis=1, inplace=True)
dataset['dst_host_srv_serror_rate'].value_counts()
dataset.drop('dst_host_srv_serror_rate', axis=1, inplace=True)
dataset['dst_host_rerror_rate'].value_counts()
dataset.drop('dst_host_rerror_rate', axis=1, inplace=True)
dataset['dst_host_srv_rerror_rate'].value_counts()
dataset.drop('dst_host_srv_rerror_rate', axis=1, inplace=True)

```

```

dataset.head()
#dataset.describe()

```

	duration	protocol_type	service	flag	src_bytes	dst_bytes	land	lnum_compromi
0	0	tcp	http	SF	181	5450	0	
1	0	tcp	http	SF	239	486	0	
2	0	tcp	http	SF	235	1337	0	
3	0	tcp	http	SF	219	1337	0	
4	0	tcp	http	SF	217	2032	0	

```
dataset['label'] = dataset['label'].replace(['back', 'buffer_overflow', 'ftp_write', 'gues
```

```
dataset.describe()
```

	duration	src_bytes	dst_bytes	land	lnum_compromised
<b>count</b>	494020.000000	4.940200e+05	4.940200e+05	494020.000000	494020.000000
<b>mean</b>	47.979400	3.025616e+03	8.685308e+02	0.000045	0.010212
<b>std</b>	707.747185	9.882191e+05	3.304003e+04	0.006673	1.798328
<b>min</b>	0.000000	0.000000e+00	0.000000e+00	0.000000	0.000000
<b>25%</b>	0.000000	4.500000e+01	0.000000e+00	0.000000	0.000000
<b>50%</b>	0.000000	5.200000e+02	0.000000e+00	0.000000	0.000000
<b>75%</b>	0.000000	1.032000e+03	0.000000e+00	0.000000	0.000000
<b>max</b>	58329.000000	6.933756e+08	5.155468e+06	1.000000	884.000000

```

x = dataset.iloc[:, :-1].values
#x
y = dataset.iloc[:, 13].values

```

```
#y
```

```
x
```

```
↳ array([[0, 'tcp', 'http', ..., 8, 9, 9],
         [0, 'tcp', 'http', ..., 8, 19, 19],
         [0, 'tcp', 'http', ..., 8, 29, 29],
         ...,
         [0, 'tcp', 'http', ..., 18, 16, 255],
         [0, 'tcp', 'http', ..., 12, 26, 255],
         [0, 'tcp', 'http', ..., 35, 6, 255]], dtype=object)
```

```
x.shape
```

```
↳ (494020, 13)
```

```
y.shape
```

```
↳ (494020,)
```

```
from sklearn.preprocessing import LabelEncoder, OneHotEncoder
from sklearn.compose import ColumnTransformer
labelencoder_x_1 = LabelEncoder()
labelencoder_x_2 = LabelEncoder()
labelencoder_x_3 = LabelEncoder()
x[:, 1] = labelencoder_x_1.fit_transform(x[:, 1])
x[:, 2] = labelencoder_x_2.fit_transform(x[:, 2])
x[:, 3] = labelencoder_x_3.fit_transform(x[:, 3])
```

```
x
```

```
↳ array([[0, 1, 22, ..., 8, 9, 9],
         [0, 1, 22, ..., 8, 19, 19],
         [0, 1, 22, ..., 8, 29, 29],
         ...,
         [0, 1, 22, ..., 18, 16, 255],
         [0, 1, 22, ..., 12, 26, 255],
         [0, 1, 22, ..., 35, 6, 255]], dtype=object)
```

```
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.3, random_state =
```

```
#feature scaling
```

```
from sklearn.preprocessing import StandardScaler
sc_x = StandardScaler()
x_train = sc_x.fit_transform(x_train)
x_test = sc_x.transform(x_test)
```

```
from sklearn.naive_bayes import GaussianNB
classifier = GaussianNB()
classifier.fit(x_train, y_train)
```

```
➞ GaussianNB(priors=None, var_smoothing=1e-09)
```

```
# Predicting the Test set results
y_pred = classifier.predict(x_test)
```

```
# Making the Confusion Matrix
from sklearn.metrics import confusion_matrix
confusion_matrix(y_test, y_pred)
```

```
➞ array([[117292, 1559],
        [ 7624, 21731]])
```

```
from sklearn.model_selection import cross_val_score
accuracies = cross_val_score(estimator = classifier, X = x_train, y = y_train, cv = 5)
accuracies.mean()
accuracies.std()
```

```
➞ 0.01824981895534346
```

```
from sklearn import metrics
# Model Accuracy, how often is the classifier correct?
print("Accuracy nb:",metrics.accuracy_score(y_test, y_pred))
```

```
➞ Accuracy nb: 0.9380389457916684
```

```
from sklearn import metrics
# Model Accuracy, how often is the classifier correct?
print(metrics.classification_report(y_test, y_pred))
```

```
➞
```

	precision	recall	f1-score	support
attack	0.94	0.99	0.96	118851
normal	0.93	0.74	0.83	29355
accuracy			0.94	148206
macro avg	0.94	0.86	0.89	148206
weighted avg	0.94	0.94	0.94	148206

```
from sklearn.ensemble import GradientBoostingClassifier
from sklearn.metrics import accuracy_score
#for optimization i used AdaBoostClassifier
```

```
gbt = GradientBoostingClassifier()
#abc2=AdaBoostClassifier(n_estimators=10,base_estimator=gbt,learning_rate=0.01)
gbt1=gbt.fit(x_train,y_train)
predictions = gbt1.predict(x_test)
print("accuracy:",accuracy_score(y_test, predictions)*100)
```

```
➞
```

```

from sklearn.linear_model import SGDClassifier
sgb = SGDClassifier(loss="hinge", penalty="l1")
#abc3=AdaBoostClassifier(n_estimators=100,base_estimator=sgb,learning_rate=0.01)
sgb1=sgb.fit(x_train, y_train)
predictions = sgb1.predict(x_test)#
print("accuracy for SGD:",accuracy_score(y_test, predictions)*100)

```

↳ accuracy for SGD: 98.5648354317639

```

from sklearn.ensemble import AdaBoostClassifier
abt = AdaBoostClassifier(n_estimators=100)

abt1=abt.fit(x_train, y_train)
predictions = abt1.predict(x_test)
print("accuracy:",accuracy_score(y_test, predictions)*100)

```

↳ accuracy: 99.8623537508603

```

from sklearn.tree import DecisionTreeClassifier
clf=DecisionTreeClassifier(max_leaf_nodes=15,criterion='gini')
#abc5=AdaBoostClassifier(n_estimators=100,base_estimator=clf,learning_rate=0.01)
clf1=clf.fit(x_train,y_train)
predictions = clf1.predict(x_test)
print("accuracy for decision tree:",accuracy_score(y_test, predictions)*100)

```

↳ accuracy for decision tree: 99.8319906076677

```

from sklearn.ensemble import RandomForestClassifier
clf2 = RandomForestClassifier(n_estimators=1000,max_leaf_nodes=15)
#abc6=AdaBoostClassifier(n_estimators=100,base_estimator=clf2,learning_rate=0.01)
clf5=clf2.fit(x_train,y_train)
predictions = clf5.predict(x_test)
print("accuracy for RFC:",accuracy_score(y_test, predictions)*100)

```

↳ accuracy for RFC: 99.72740644778214

```

from sklearn.neural_network import MLPClassifier
mlp = MLPClassifier(hidden_layer_sizes=(100, 100, 12), alpha=1e-4, solver='sgd', random_st
model_3 = mlp.fit(x_train, y_train)
y_pred = model_3.predict(x_test)
acc = metrics.accuracy_score(y_test, y_pred)
print("This is Multi Layer Perceptron classifier \n\n")
print("Accuracy: {:.4f} %".format(acc*100))

```

↳ This is Multi Layer Perceptron classifier

Accuracy: 99.8576 %

