

FeatureSelection

March 13, 2023

```
[4]: %matplotlib inline

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

import warnings
warnings.filterwarnings('ignore')
pd.options.display.max_columns = None

from sklearn.ensemble import RandomForestClassifier
from sklearn.model_selection import StratifiedKFold
from sklearn.feature_selection import RFECV

from numpy import mean
from numpy import std
from sklearn.datasets import make_classification
from sklearn.model_selection import cross_val_score
from sklearn.model_selection import RepeatedKFold
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestRegressor
from sklearn.pipeline import Pipeline
from sklearn import preprocessing
from sklearn.preprocessing import MinMaxScaler
```

```
[78]: #read dataset
data = pd.read_csv("/content/drive/MyDrive/CIND 820 Capstone Project/
↳merged_completedata.csv")
```

```
[79]: #checking dimensions of data
data.head()
```

```
[79]:
```

	RecordID	X	Y	FID	BusinessID	\
0	1	-79.689829	43.644181	1	1055	
1	2	-79.689419	43.644988	2	1057	
2	3	-79.689419	43.644988	3	1058	
3	4	-79.689419	43.644988	4	1060	

4 5 -79.690664 43.645493 5 1061

	Name	Address	StreetNo	\
0	Golf Trends Inc.	300 Ambassador Dr	300	
1	Apex Graphics Inc.	320 Ambassador Dr	320	
2	Sands, John & Associates Limited	320 Ambassador Dr	320	
3	Printmedia-Tackaberry Times	320 Ambassador Dr	320	
4	S W R Industries Ltd.	321 Ambassador Dr	321	

	StreetName	BldgNo	UnitNo	PostalCode	Location	Ward	NAICSCode	\
0	Ambassador Dr	No	No	L5T	Gateway EA (East)	5	41	
1	Ambassador Dr	No	No	L5T	Gateway EA (East)	5	32	
2	Ambassador Dr	No	No	L5T	Gateway EA (East)	5	32	
3	Ambassador Dr	No	No	L5T	Gateway EA (East)	5	32	
4	Ambassador Dr	No	No	L5T	Gateway EA (East)	5	41	

	NAICSCat	NAICSDescr	\
0	Wholesale Trade	Amusement and Sporting Goods Wholesaler-Distri...	
1	Manufacturing	Support Activities for Printing	
2	Manufacturing	Support Activities for Printing	
3	Manufacturing	Other Printing	
4	Wholesale Trade	Industrial Machinery, Equipment and Supplies W...	

	Phone	Fax	TollFree	EMail	WebAddress	EmplRange	\
0	905-795-8900	905-795-8988	Yes	Yes	Yes	3	
1	905-795-9575	905-795-8775	No	Yes	Yes	4	
2	905-795-9519	905-795-8775	No	No	No	5	
3	905-564-8121	905-564-7395	No	Yes	Yes	1	
4	905-564-8080	905-564-5003	No	Yes	Yes	2	

	CENT_X	CENT_Y	Year	isnew	Closed
0	605668.2538	4.833187e+06	2016	No	No
1	605699.9370	4.833277e+06	2016	No	No
2	605699.9370	4.833277e+06	2016	No	No
3	605699.9370	4.833277e+06	2016	No	No
4	605598.6442	4.833332e+06	2016	No	No

```
[80]: #describe categorical data
data.describe(include='O')
```

```
[80]:
```

	Name	Address	StreetName	BldgNo	UnitNo	PostalCode	\
count	78032	78032	78032	78032	78032	78032	
unique	22710	6618	669	2	2	37	
top	Subway	100 City Centre Dr	Dundas St E	No	Yes	L4W	
freq	212	953	3202	73798	53665	12410	

	Location	NAICSCat	NAICSDescr	\
--	----------	----------	------------	---

count	78032	78032	78032
unique	56	19	1039
top	Northeast EA (West)	Retail Trade	Limited-service eating places
freq	21104	11071	3647

	Phone	Fax	TollFree	EMail	WebAddress	isnew	Closed
count	78032	78032	78032	78032	78032	78032	78032
unique	25064	15752	2	2	2	2	2
top			No	Yes	Yes	No	No
freq	1457	29473	66596	47406	56765	71148	71617

```
[81]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 78032 entries, 0 to 78031
Data columns (total 28 columns):
#   Column          Non-Null Count  Dtype
---  -
0   RecordID        78032 non-null  int64
1   X               78032 non-null  float64
2   Y               78032 non-null  float64
3   FID             78032 non-null  int64
4   BusinessID      78032 non-null  int64
5   Name            78032 non-null  object
6   Address         78032 non-null  object
7   StreetNo        78032 non-null  int64
8   StreetName      78032 non-null  object
9   BldgNo          78032 non-null  object
10  UnitNo          78032 non-null  object
11  PostalCode      78032 non-null  object
12  Location        78032 non-null  object
13  Ward            78032 non-null  int64
14  NAICSCode       78032 non-null  int64
15  NAICSCat        78032 non-null  object
16  NAICSDescr      78032 non-null  object
17  Phone           78032 non-null  object
18  Fax             78032 non-null  object
19  TollFree        78032 non-null  object
20  EMail           78032 non-null  object
21  WebAddress      78032 non-null  object
22  EmplRange       78032 non-null  int64
23  CENT_X          78032 non-null  float64
24  CENT_Y          78032 non-null  float64
25  Year            78032 non-null  int64
26  isnew           78032 non-null  object
27  Closed          78032 non-null  object
dtypes: float64(4), int64(8), object(16)
```

memory usage: 16.7+ MB

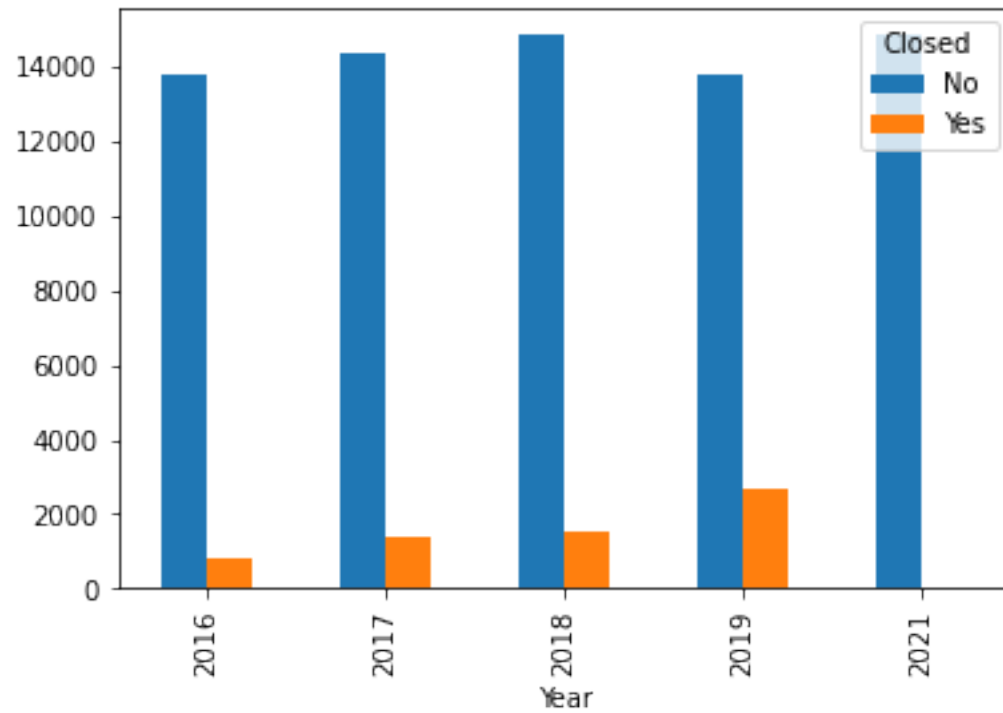
```
[82]: #NAICSCode back to object
data['NAICSCode'] = data['NAICSCode'].astype(str)
```

```
[83]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 78032 entries, 0 to 78031
Data columns (total 28 columns):
 #   Column          Non-Null Count  Dtype
---  -
 0   RecordID        78032 non-null  int64
 1   X                78032 non-null  float64
 2   Y                78032 non-null  float64
 3   FID              78032 non-null  int64
 4   BusinessID      78032 non-null  int64
 5   Name             78032 non-null  object
 6   Address          78032 non-null  object
 7   StreetNo        78032 non-null  int64
 8   StreetName      78032 non-null  object
 9   BldgNo          78032 non-null  object
10   UnitNo          78032 non-null  object
11   PostalCode      78032 non-null  object
12   Location        78032 non-null  object
13   Ward            78032 non-null  int64
14   NAICSCode       78032 non-null  object
15   NAICSCat        78032 non-null  object
16   NAICSDescr      78032 non-null  object
17   Phone           78032 non-null  object
18   Fax             78032 non-null  object
19   TollFree        78032 non-null  object
20   EMail           78032 non-null  object
21   WebAddress      78032 non-null  object
22   EmplRange       78032 non-null  int64
23   CENT_X          78032 non-null  float64
24   CENT_Y          78032 non-null  float64
25   Year            78032 non-null  int64
26   isNew           78032 non-null  object
27   Closed          78032 non-null  object
dtypes: float64(4), int64(7), object(17)
memory usage: 16.7+ MB
```

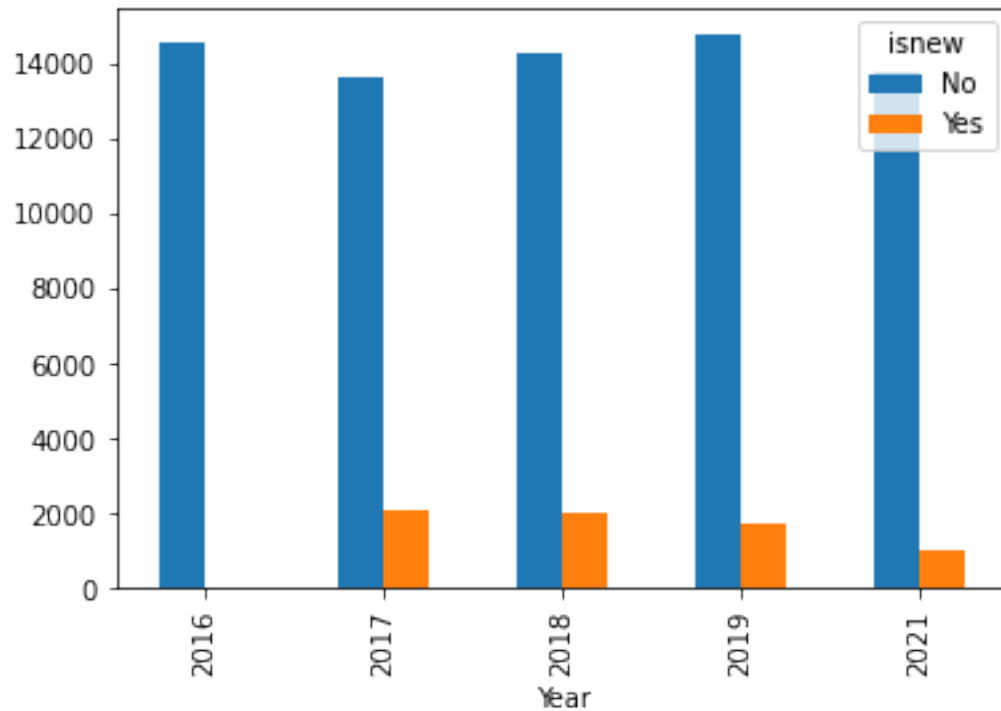
```
[84]: df_gb_openclosed = data.groupby(['Year', 'Closed']).size().unstack(level=1)
df_gb_openclosed.plot(kind = 'bar')
```

```
[84]: <AxesSubplot:xlabel='Year'>
```



```
[11]: df_gb_openclosed = data.groupby(['Year', 'isnew']).size().unstack(level=1)
df_gb_openclosed.plot(kind = 'bar')
```

```
[11]: <AxesSubplot:xlabel='Year'>
```



```
[85]: data.head()
```

```
[85]:
```

	RecordID	X	Y	FID	BusinessID	\
0	1	-79.689829	43.644181	1	1055	
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	NAICSCat	NAICSDescr \
0	Wholesale Trade	Amusement and Sporting Goods Wholesaler-Distri...
1	Manufacturing	Support Activities for Printing
2	Manufacturing	Support Activities for Printing
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	Phone	Fax	TollFree	EMail	WebAddress	EmplRange \
0	905-795-8900	905-795-8988	Yes	Yes	Yes	3
1	905-795-9575	905-795-8775	No	Yes	Yes	4
2	905-795-9519	905-795-8775	No	No	No	5
3	905-564-8121	905-564-7395	No	Yes	Yes	1
4	905-564-8080	905-564-5003	No	Yes	Yes	2

	CENT_X	CENT_Y	Year	isnew	Closed
0	605668.2538	4.833187e+06	2016	No	No
1	605699.9370	4.833277e+06	2016	No	No
2	605699.9370	4.833277e+06	2016	No	No
3	605699.9370	4.833277e+06	2016	No	No
4	605598.6442	4.833332e+06	2016	No	No

```
[86]: #describe categorical data
data.describe(include='O')
```

```
[86]:
```

	Name	Address	StreetName	BldgNo	UnitNo	PostalCode \
count	78032	78032	78032	78032	78032	78032
unique	22710	6618	669	2	2	37
top	Subway	100 City Centre Dr	Dundas St E	No	Yes	L4W
freq	212	953	3202	73798	53665	12410

	Location	NAICSCode	NAICSCat \
count	78032	78032	78032
unique	56	24	19
top	Northeast EA (West)	81	Retail Trade
freq	21104	9052	11071

	NAICSDescr	Phone	Fax	TollFree	EMail \
count	78032	78032	78032	78032	78032
unique	1039	25064	15752	2	2
top	Limited-service eating places			No	Yes
freq	3647	1457	29473	66596	47406

	WebAddress	isnew	Closed
count	78032	78032	78032
unique	2	2	2
top	Yes	No	No
freq	56765	71148	71617

```
[87]: #drop columns that have unique values and categorical
data.drop(['FID','BusinessID','RecordID', 'Name','StreetNo','Address',
↳'NAICSCat',
↳'StreetName','Location','Phone','Fax','NAICSDescr','EMail','PostalCode','BldgNo','UnitNo','
↳'NAICSCode'], axis=1, inplace=True)
```

```
[88]: # Save the new data set to a new file
data.to_csv("/content/drive/MyDrive/CIND 820 Capstone Project/
↳categoricaltonumericdata.csv", index=False)
```

```
[89]: data = data[data['Year'] == 2019]
```

```
[90]: data.head()
```

```
[90]:
```

	X	Y	Ward	EmplRange	CENT_X	CENT_Y	Year	\
46689	-79.665386	43.684736	5	1	607567.2334	4.837723e+06	2019	
46690	-79.642760	43.593515	4	2	609556.5032	4.827621e+06	2019	
46691	-79.667311	43.682752	5	3	607415.6044	4.837500e+06	2019	
46692	-79.629235	43.698932	4	2	610454.8654	4.839347e+06	2019	
46693	-79.629235	43.698932	4	4	610454.8654	4.839347e+06	2019	

```

Closed
46689    No
46690    No
46691    No
46692    No
46693    No
```

```
[ ]: #data = data[data['Closed'] == 0]
#use this for when taking 2021 and is new!!!
```

```
[63]: data.head()
```

```
[63]:
```

	X	Y	StreetNo	Ward	EmplRange	CENT_X	\
46689	-79.665386	43.684736	7050	5	1	607567.2334	
46690	-79.642760	43.593515	100	4	2	609556.5032	
46691	-79.667311	43.682752	7015	5	3	607415.6044	
46692	-79.629235	43.698932	100	4	2	610454.8654	
46693	-79.629235	43.698932	100	4	4	610454.8654	

```

CENT_Y  Year  Closed
46689  4.837723e+06  2019    No
46690  4.827621e+06  2019    No
46691  4.837500e+06  2019    No
46692  4.839347e+06  2019    No
46693  4.839347e+06  2019    No
```



```
[91]: df2 = data.mean(axis=0)
      print(df2)
```

```
X          -7.965769e+01
Y           4.360136e+01
Ward         5.372927e+00
EmplRange    2.183981e+00
CENT_X       6.088039e+05
CENT_Y       4.828662e+06
Year         2.019000e+03
dtype: float64
```

```
[92]: correlated_features= set()
      correlation_matrix = data.drop('Closed', axis=1).corr()

      for i in range(len(correlation_matrix.columns)):
          for j in range(i):
              if abs(correlation_matrix.iloc[i,j]) > 0.8:
                  colname = correlation_matrix.columns[i]
                  correlated_features.add(colname)
```

```
[93]: correlated_features
```

```
[93]: {'CENT_X', 'CENT_Y'}
```

```
[94]: data.drop(['CENT_X', 'CENT_Y'], axis=1, inplace=True)
      #use for closed analysys
```

```
[95]: #checking dimensions of data
      data.head()
```

```
[95]:
```

	X	Y	Ward	EmplRange	Year	Closed
46689	-79.665386	43.684736	5	1	2019	No
46690	-79.642760	43.593515	4	2	2019	No
46691	-79.667311	43.682752	5	3	2019	No
46692	-79.629235	43.698932	4	2	2019	No
46693	-79.629235	43.698932	4	4	2019	No

```
[96]: data.dtypes
```

```
[96]: X          float64
      Y          float64
      Ward       int64
      EmplRange  int64
      Year       int64
      Closed     object
      dtype: object
```

```
[97]: data['EmplRange'].isnull().values.any()
```

```
[97]: False
```

```
[98]: data.isnull().sum().sum()
```

```
[98]: 0
```

```
[116]: X = data.drop('Closed', axis=1)
target = data['Closed']

rfc = RandomForestClassifier(random_state=101)
rfecv = RFECV(estimator=rfc, step=1, cv=StratifiedKFold(10), scoring='accuracy')
rfecv.fit(X, target)
#Recursive feature elimination
#Takes around 2-3 minutes to run. Not as effecient for feature selection.
```

```
[116]: RFECV(cv=StratifiedKFold(n_splits=10, random_state=None, shuffle=False),
          estimator=RandomForestClassifier(random_state=101), scoring='accuracy')
```

```
[100]: print('Optimal number of features: {}'.format(rfecv.n_features_))
```

```
Optimal number of features: 3
```

```
[101]: print(np.where(rfecv.support_ == False)[0])

X.drop(X.columns[np.where(rfecv.support_ == False)[0]], axis=1, inplace=True)
```

```
[2 4]
```

```
[102]: rfecv.estimator_.feature_importances_
```

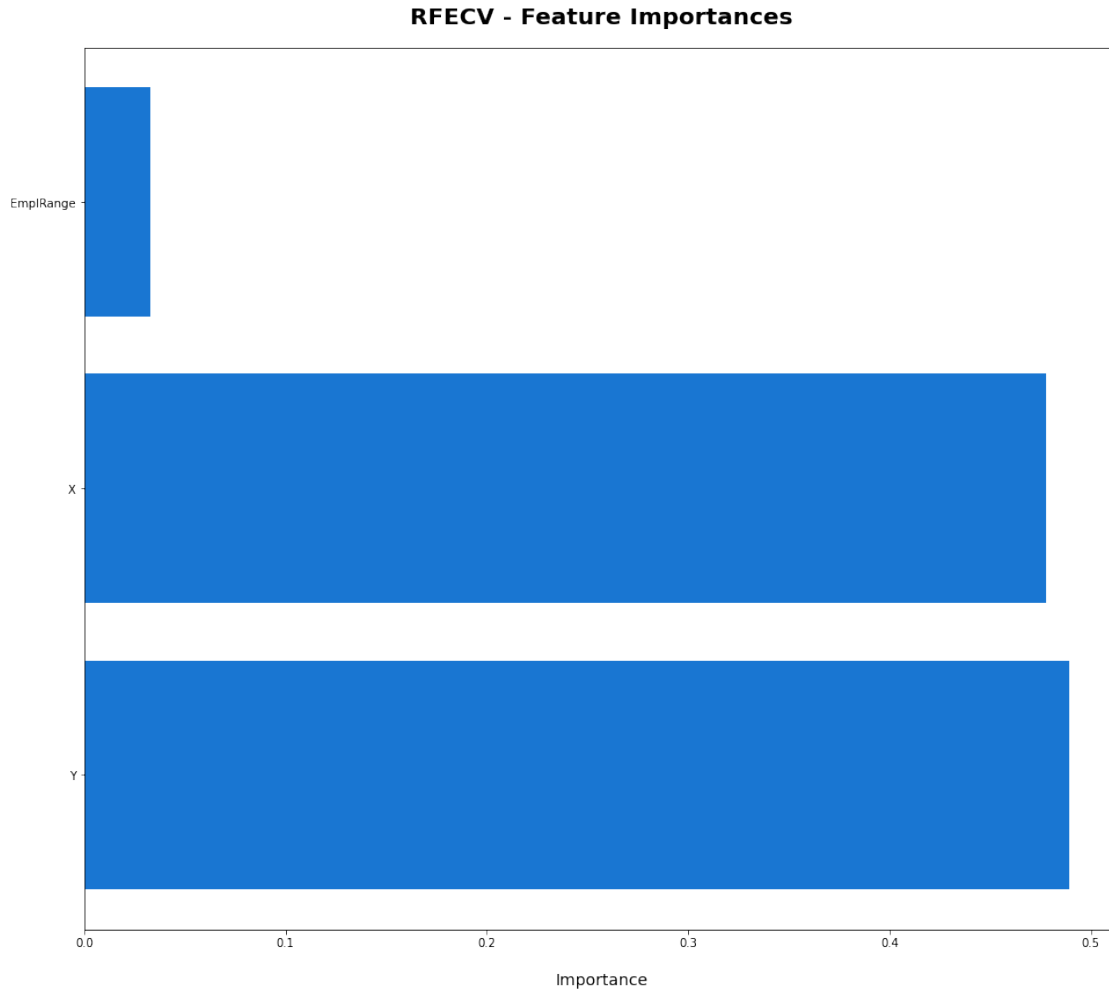
```
[102]: array([0.47756572, 0.48930314, 0.03313114])
```

```
[115]: dset = pd.DataFrame()
dset['attr'] = X.columns
dset['importance'] = rfecv.estimator_.feature_importances_

dset = dset.sort_values(by='importance', ascending=False)

plt.figure(figsize=(16, 14))
plt.barh(y=dset['attr'], width=dset['importance'], color='#1976D2')
plt.title('RFECV - Feature Importances', fontsize=20, fontweight='bold', pad=20)
plt.xlabel('Importance', fontsize=14, labelpad=20)
#plt.show()
```

```
[115]: Text(0.5, 0, 'Importance')
```



```
[ ]: plt.figure(figsize=(16, 9))
plt.title('Recursive Feature Elimination with Cross-Validation', fontsize=18,
         fontweight='bold', pad=20)
plt.xlabel('Number of features selected', fontsize=14, labelpad=20)
plt.ylabel('% Correct Classification', fontsize=14, labelpad=20)
plt.plot(range(1, len(rfecv.grid_scores_) + 1), rfecv.grid_scores_,
         color='#303F9F', linewidth=3)

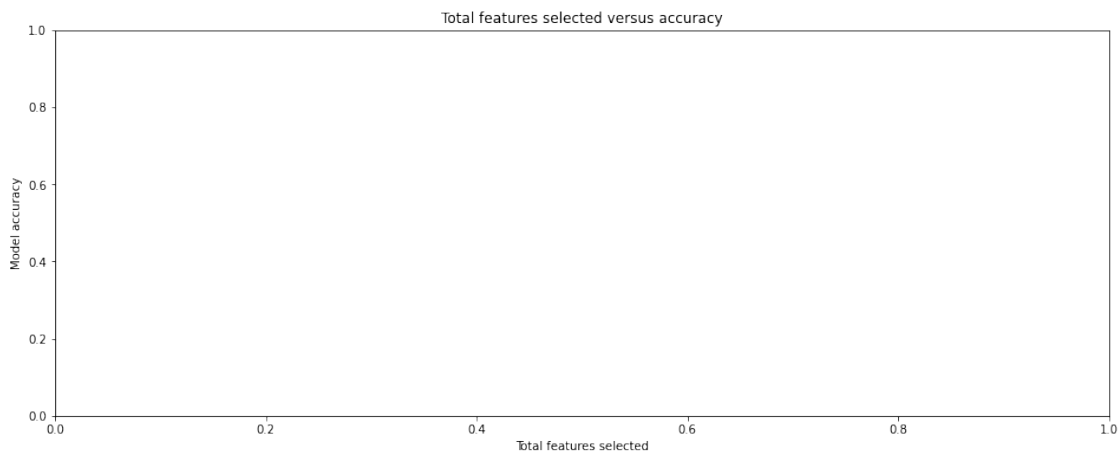
plt.show()
```

```
[110]: plt.figure( figsize=(16, 6))
plt.title('Total features selected versus accuracy')
plt.xlabel('Total features selected')
plt.ylabel('Model accuracy')
```

```
plt.plot(range(1, len(rfecv.grid_scores_) + 1), rfecv.grid_scores_)
plt.show()
```

```
-----
AttributeError                                Traceback (most recent call last)
<ipython-input-110-31a34b31d8b7> in <module>
      3 plt.xlabel('Total features selected')
      4 plt.ylabel('Model accuracy')
----> 5 plt.plot(range(1, len(rfecv.grid_scores_) + 1), rfecv.grid_scores_)
      6 plt.show()

AttributeError: 'RFECV' object has no attribute 'grid_scores_'
```



```
[111]: df_features = pd.DataFrame(columns = ['feature', 'support', 'ranking'])

for i in range(X.shape[1]):
    row = {'feature': i, 'support': rfecv.support_[i], 'ranking': rfecv.
    ranking_[i]}
    df_features = df_features.append(row, ignore_index=True)

df_features.sort_values(by='ranking').head(10)
```

```
[111]:  feature support ranking
0         0     True         1
1         1     True         1
2         2    False         2
```

```
[112]: df_features[df_features['support']==True]
```

```
[112]: feature support ranking
      0      0      True      1
      1      1      True      1
```

```
[114]: selected_features = rfecv.get_support(1)
      X = data[data.columns[selected_features]]
```