Out[155]:

	Cereal
0	Manufacturer
1	Туре
2	Calories
3	Protein
4	Fat
5	Sodium
6	Dietary Fiber
7	Complex Carbohydrates
8	Sugar
9	Display Shelf
10	Potassium
11	Vitamins & Minerals
12	Weight

```
In [156]:
          col_names_list=[]
          col_names_list.append("Cereal")
          col_names_list
Out[156]: ['Cereal']
In [157]: col_names_list2=col_names["Cereal"].tolist()
In [158]: for i in range(len(col_names_list2)):
               col_names_list.append(col_names_list2[i])
          col_names_list
Out[158]: ['Cereal',
            'Manufacturer',
            'Type',
            'Calories',
            'Protein',
            'Fat',
            'Sodium',
            'Dietary Fiber',
            'Complex Carbohydrates',
            'Sugar ',
            'Display Shelf',
            'Potassium',
            'Vitamins & Minerals',
            'Weight']
```

```
In [159]:
          counter=0
          main dic={}
          for col in col_names_list:
              main_dic[col]=[]
          main dic
Out[159]: {'Cereal': [],
            'Manufacturer': [],
            'Type': [],
            'Calories': [],
            'Protein': [],
            'Fat': [],
            'Sodium': [],
            'Dietary Fiber': [],
            'Complex Carbohydrates': [],
            'Sugar ': [],
            'Display Shelf': [],
            'Potassium': [],
            'Vitamins & Minerals': [],
            'Weight': []}
In [160]: for line in df:
               values=line.strip().split(" ")
               for i in range(len(col_names_list)):
                   main_dic[col_names_list[i]].append(values[i])
               counter+=1
           counter
Out[160]: 77
```

```
In [161]: main_dic
              Appre_crimamon_cheer.ros ,
             'Apple_Jacks',
             'Basic_4',
             'Bran_Chex',
             'Bran Flakes',
             "Cap'n'Crunch",
             'Cheerios',
             'Cinnamon_Toast_Crunch',
             'Clusters',
             'Cocoa_Puffs',
             'Corn_Chex',
             'Corn_Flakes',
             'Corn_Pops',
             'Count Chocula',
             "Cracklin'_Oat_Bran",
             'Cream_of_Wheat_(Quick)',
             'Crispix',
             'Crispy_Wheat_&_Raisins',
             'Double_Chex',
             'Froot_Loops',
In [162]: data=pd.DataFrame(main_dic)
```

In [163]: data.head()

Out[163]:

	Cereal	Manufacturer	Туре	Calories	Protein	Fat	Sodium	Dietary Fiber	Complex Carbohydrates	Sugar	Display Shelf	Potassium	Vitamins & Minerals
0	100%_Bran	N	С	70	4	1	130	10	5	6	3	280	2!
1	100%_Natural_Bran	Q	С	120	3	5	15	2	8	8	3	135	(
2	All-Bran	K	С	70	4	1	260	9	7	5	3	320	2!
3	All- Bran_with_Extra_Fiber	К	С	50	4	0	140	14	8	0	3	330	2!
4	Almond_Delight	R	С	110	2	2	200	1	14	8	3	-1	2!
4													>

```
In [185]: data.shape
Out[185]: (77, 15)
          data[["Calories", "Protein", "Fat", "Sodium", "Dietary Fiber", "Complex Carbohydrates", "Sugar ", "Potassium", "\
In [164]:
           data.dtypes
Out[164]: Cereal
                                     object
          Manufacturer
                                     object
          Type
                                     object
          Calories
                                      int64
          Protein
                                      int64
          Fat
                                      int64
          Sodium
                                      int64
          Dietary Fiber
                                    float64
          Complex Carbohydrates
                                    float64
          Sugar
                                      int64
          Display Shelf
                                     object
          Potassium
                                      int64
          Vitamins & Minerals
                                      int64
          Weight
                                     object
          dtype: object
          from sklearn.linear model import LinearRegression
In [165]:
          from sklearn.feature selection import RFE
          from sklearn.svm import SVR
In [166]: feature cols=["Protein", "Fat", "Sodium", "Dietary Fiber", "Complex Carbohydrates", "Sugar ", "Potassium", "Vitar
```

```
col names=data.columns.values.tolist()
In [167]:
           col names
Out[167]: ['Cereal',
            'Manufacturer',
            'Type',
            'Calories',
            'Protein',
            'Fat',
            'Sodium',
            'Dietary Fiber',
            'Complex Carbohydrates',
            'Sugar ',
            'Display Shelf',
            'Potassium',
            'Vitamins & Minerals',
            'Weight']
In [168]:
          X=data[feature_cols]
          Y=data["Calories"]
          estimator=SVR(kernel="linear")
In [169]:
          selector=RFE(estimator, 8, step=1)
          selector=selector.fit(X,Y)
In [170]: | selector.support
Out[170]: array([ True, True, True, True, True, True, True])
In [171]: | selector.ranking_
Out[171]: array([1, 1, 1, 1, 1, 1, 1, 1])
In [172]: lm=LinearRegression()
          lm.fit(X,Y)
Out[172]: LinearRegression(copy_X=True, fit_intercept=True, n_jobs=None,
                   normalize=False)
```

```
In [173]:
          print(lm.intercept )
          print(lm.coef )
          22.11145682760973
          [ 5.74639942e+00 8.43287251e+00 2.45646695e-03 -9.39108086e-01
            2.81413513e+00 3.22898332e+00 -6.64605092e-03 1.55397354e-02
In [174]: lm.score(X,Y)
Out[174]: 0.8357452532148784
In [175]: list(zip(feature cols,lm.coef ))
Out[175]: [('Protein', 5.746399418090031),
           ('Fat', 8.432872512221525),
           ('Sodium', 0.0024564669528756933),
           ('Dietary Fiber', -0.9391080862390954),
           ('Complex Carbohydrates', 2.814135127424553),
           ('Sugar', 3.228983316982189),
           ('Potassium', -0.006646050916592121),
           ('Vitamins & Minerals', 0.015539735440727198)]
          calories pred=lm.predict(X)
In [176]:
```

```
In [191]: data["Calories_Pred"]=calories_pred
    data.head(10)
```

Out[191]:

	Cornel	Manufacturer	Tuna	Calories	Protein	Fat	Sodium	Dietary Fiber	Complex Carbohydrates	Sugar	Display Shelf	Potassium	Vita
	Cereal	Manufacturer	Type										Min
0	100%_Bran	N	С	70	4	1	130	10.0	5.0	6	3	280	
1	100%_Natural_Bran	Q	С	120	3	5	15	2.0	8.0	8	3	135	
2	All-Bran	K	С	70	4	1	260	9.0	7.0	5	3	320	
3	All-Bran_with_Extra_Fiber	K	С	50	4	0	140	14.0	8.0	0	3	330	
4	Almond_Delight	R	С	110	2	2	200	1.0	14.0	8	3	-1	
5	Apple_Cinnamon_Cheerios	G	С	110	2	2	180	1.5	10.5	10	1	70	
6	Apple_Jacks	K	С	110	2	0	125	1.0	11.0	14	2	30	
7	Basic_4	G	С	130	3	2	210	2.0	18.0	8	3	100	
8	Bran_Chex	R	С	90	2	1	200	4.0	15.0	6	1	125	
9	Bran_Flakes	Р	С	90	3	0	210	5.0	13.0	5	3	190	

•

```
In [178]: calories_mean=np.mean(data["Calories"])
  calories_mean
```

Out[178]: 106.88311688311688

In [179]: SSD=sum((data["Calories"]-data["Calories_Pred"])**2)
SSD

Out[179]: 4739.069421530211

```
In [180]: RSE=np.sqrt(SSD/(len(data)-len(feature_cols)-1))
    RSE
Out[180]: 8.348185274384145

In [184]: error=RSE/calories_mean
    error
Out[184]: 0.07810574315037414

In []:
```