

In []:

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

In [82]:

```
dataset = pd.read_csv("https://archive.ics.uci.edu/ml/machine-learning-database
s/glass/glass.data")
```

```
dataset.columns=["No.", "RI", "Na", "Mg", "Al", "Si", "K", "Ca", "Ba", "Fe", "Type"]
```

In [83]:

```
dataset=dataset.drop("No.",axis=1)
```

In [84]:

```
dataset.head()
```

Out[84]:

	RI	Na	Mg	Al	Si	K	Ca	Ba	Fe	Type
0	1.51761	13.89	3.60	1.36	72.73	0.48	7.83	0.0	0.00	1
1	1.51618	13.53	3.55	1.54	72.99	0.39	7.78	0.0	0.00	1
2	1.51766	13.21	3.69	1.29	72.61	0.57	8.22	0.0	0.00	1
3	1.51742	13.27	3.62	1.24	73.08	0.55	8.07	0.0	0.00	1
4	1.51596	12.79	3.61	1.62	72.97	0.64	8.07	0.0	0.26	1

In [85]:

```
print(dataset.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 213 entries, 0 to 212
Data columns (total 10 columns):
RI          213 non-null float64
Na          213 non-null float64
Mg          213 non-null float64
Al          213 non-null float64
Si          213 non-null float64
K           213 non-null float64
Ca          213 non-null float64
Ba          213 non-null float64
Fe          213 non-null float64
Type       213 non-null int64
dtypes: float64(9), int64(1)
memory usage: 16.7 KB
None
```

In [86]:

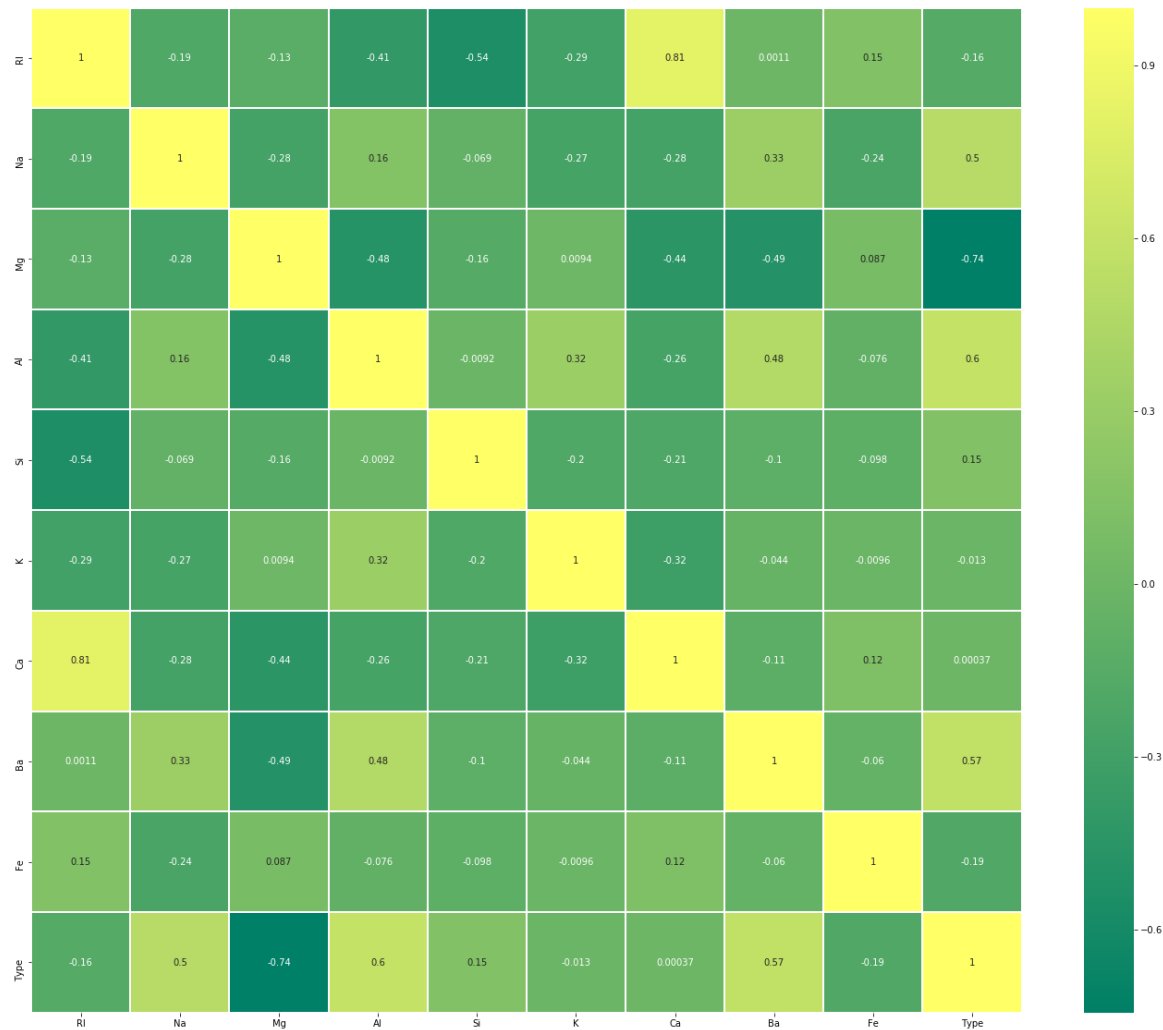
```
dataset.describe()
```

Out[86]:

	RI	Na	Mg	Al	Si	K	Ca
count	213.000000	213.000000	213.000000	213.000000	213.000000	213.000000	213.000000
mean	1.518353	13.406761	2.676056	1.446526	72.655023	0.499108	8.957934
std	0.003039	0.818371	1.440453	0.499882	0.774052	0.653035	1.426435
min	1.511150	10.730000	0.000000	0.290000	69.810000	0.000000	5.430000
25%	1.516520	12.900000	2.090000	1.190000	72.280000	0.130000	8.240000
50%	1.517680	13.300000	3.480000	1.360000	72.790000	0.560000	8.600000
75%	1.519150	13.830000	3.600000	1.630000	73.090000	0.610000	9.180000
max	1.533930	17.380000	3.980000	3.500000	75.410000	6.210000	16.190000

In [87]:

```
plt.figure(figsize=(24,20))
sns.heatmap(dataset.corr(),annot=True,linecolor="white",linewidths=(1,1),cmap="summer")
plt.show()
```



In [183]:

```
X = dataset.iloc[:, [1,2,3,4,7]].values
y = dataset.iloc[:, 9].values
```

In [184]:

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.25, random_state = 0)
```

In [185]:

```
from sklearn.neighbors import KNeighborsClassifier
classifier = KNeighborsClassifier(n_neighbors = 2, metric = 'minkowski', p = 2)
classifier.fit(X_train, y_train)
```

Out[185]:

```
KNeighborsClassifier(algorithm='auto', leaf_size=30, metric='minkows
ki',
                    metric_params=None, n_jobs=None, n_neighbors=2, p=2,
                    weights='uniform')
```

In [186]:

```
y_pred = classifier.predict(X_test)
```

In [187]:

```
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test, y_pred)
cm
```

Out[187]:

```
array([[12,  2,  0,  0,  0,  0],
       [ 9, 13,  0,  1,  0,  0],
       [ 4,  2,  0,  0,  0,  0],
       [ 0,  1,  0,  1,  0,  0],
       [ 0,  0,  0,  0,  2,  0],
       [ 0,  0,  0,  0,  0,  7]])
```

In [188]:

```
from matplotlib.colors import ListedColormap
X_set, y_set = X_train, y_train
X1, X2 = np.meshgrid(np.arange(start = X_set[:, 0].min() - 1, stop = X_set[:, 0]
                             .max() + 1, step = 0.01),
                     np.arange(start = X_set[:, 1].min() - 1, stop = X_set[:, 1]
                             .max() + 1, step = 0.01))

plt.xlim(X1.min(), X1.max())#Setting limits turns autoscaling off for the x-axis.
plt.ylim(X2.min(), X2.max())
#enumerate allows us to loop over something and have an automatic counter.
#np.unique returns the unique values from the input array
for i, j in enumerate(np.unique(y_set)):
    plt.scatter(X_set[y_set == j, 0], X_set[y_set == j, 1],
                c = ListedColormap(('red', 'green', 'b', 'orange', 'y', 'pink'))(i),
                label = j)
plt.title('K-NN (Training set)')
plt.xlabel('height')
plt.ylabel('weight')
plt.legend()
plt.show()
```

'c' argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with 'x' & 'y'. Please use a 2-D array with a single row if you really want to specify the same RGB or RGBA value for all points.

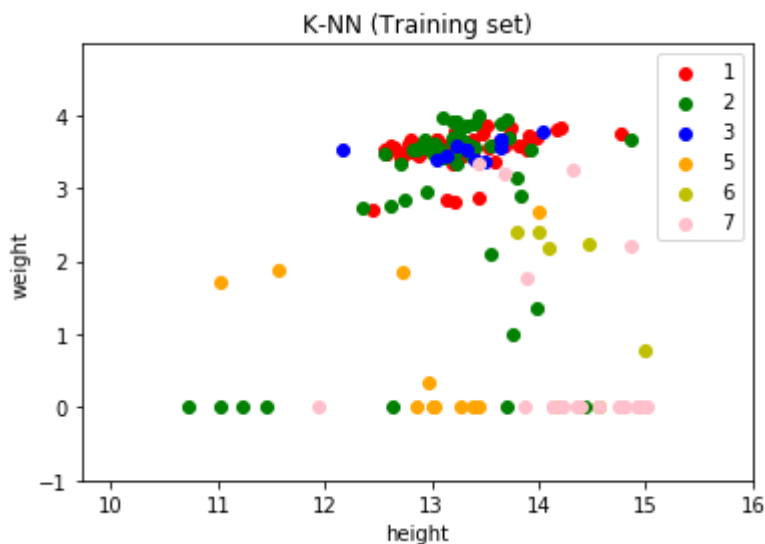
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In [189]:

```
from sklearn.model_selection import cross_val_score
from sklearn.neighbors import KNeighborsClassifier
scores=[]
for k in range(1,32):
    knn=KNeighborsClassifier(k)
    score_val=cross_val_score(knn,X_train,y_train,scoring='accuracy',cv=10)
    score_mean=score_val.mean()
    scores.append(score_mean)

# get index of maxium score along axis, default axis=0 for 1 dimensional array
best_k=np.argmax(scores)+1
print(best_k)
# generate KNN model
knn=KNeighborsClassifier(best_k)
# fit with train data set
knn.fit(X_train,y_train)
# get Modes presicion rate using test set
print("prediction precision rate:",knn.score(X_test,y_test))
```

8/14


```
odel_selection/_split.py:652: Warning: The least populated class in
y has only 7 members, which is too few. The minimum number of member
s in any class cannot be less than n_splits=10.
% (min_groups, self.n_splits)), Warning)
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```

```
% (min_groups, self.n_splits)), Warning)
```

```
2
```

```
prediction precision rate: 0.6481481481481481
```

```
/Users/nandanadileep/anaconda3/lib/python3.7/site-packages/sklearn/model_selection/_split.py:652: Warning: The least populated class in y has only 7 members, which is too few. The minimum number of members in any class cannot be less than n_splits=10.
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```

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% (min_groups, self.n_splits)), Warning)
```

In [190]:

```
#to increase the accuracy rate, we over sample the skewed data

df3=dataset[dataset['Type']==3]
df3=pd.concat([df3]*4)
df5=dataset[dataset['Type']==5]
df5=pd.concat([df5]*5)
df6=dataset[dataset['Type']==6]
df6=pd.concat([df6]*7)
df7=dataset[dataset['Type']==7]
df7=pd.concat([df7]*2)
df1=dataset[dataset['Type']==1]
df2=dataset[dataset['Type']==2]
df_balanced=pd.concat([df1,df2,df3,df5,df6,df7])
df_balanced.shape

dataset.head()
type=df_balanced['Type'].groupby(df_balanced['Type']).count()
type
df6
```

Out[190]:

	RI	Na	Mg	Al	Si	K	Ca	Ba	Fe	Type
175	1.51905	14.00	2.39	1.56	72.37	0.0	9.57	0.0	0.0	6
176	1.51937	13.79	2.41	1.19	72.76	0.0	9.77	0.0	0.0	6
177	1.51829	14.46	2.24	1.62	72.38	0.0	9.26	0.0	0.0	6
178	1.51852	14.09	2.19	1.66	72.67	0.0	9.32	0.0	0.0	6
179	1.51299	14.40	1.74	1.54	74.55	0.0	7.59	0.0	0.0	6
180	1.51888	14.99	0.78	1.74	72.50	0.0	9.95	0.0	0.0	6
181	1.51916	14.15	0.00	2.09	72.74	0.0	10.88	0.0	0.0	6
182	1.51969	14.56	0.00	0.56	73.48	0.0	11.22	0.0	0.0	6
183	1.51115	17.38	0.00	0.34	75.41	0.0	6.65	0.0	0.0	6
175	1.51905	14.00	2.39	1.56	72.37	0.0	9.57	0.0	0.0	6
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63 rows × 10 columns

In [191]:

```
from sklearn.model_selection import train_test_split
from sklearn.model_selection import cross_val_score
from sklearn.neighbors import KNeighborsClassifier

# df.columns is column labels property
features=df_balanced.columns[:-1].tolist()
X=df_balanced[features].values
y=df_balanced['Type']

X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.2,random_state=1)
score=[]
for i in range(32):
    knn=KNeighborsClassifier(k)
    score_val=cross_val_score(knn,X_train,y_train,scoring='accuracy',cv=10)
    score_mean=score_val.mean()
    scores.append(score_mean)
best_K=np.argmax(scores)+1
print('best K is:',best_K)
knn=KNeighborsClassifier(best_K)
knn.fit(X_train,y_train)
print("prediction precision rate:",knn.score(X_test,y_test))
```

best K is: 2
prediction precision rate: 0.8375

In []:

In []: