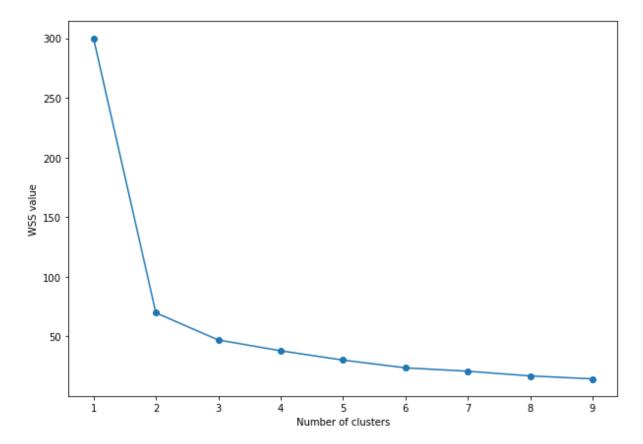
```
# -*- coding: utf-8 -*-
Created on Sat Aug 15 11:10:14 2020
K-means using euclidean distance for
@author: Walter
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sn
%matplotlib inline
data = pd.read_csv('sample.csv')
from sklearn.cluster import KMeans
from sklearn.preprocessing import StandardScaler
#data.head(10)
#data.info
X = data
X =X.values
#X
scaler = StandardScaler()
X_scaled = scaler.fit_transform( X )
```

#scale the values so one doesn't dominate

```
#X
cluster_range = range( 1, 10)
# searching for optimal K value across the range 1 - 10
cluster_errors = []
for num_clusters in cluster_range:
 clusters = KMeans( num_clusters )
 clusters.fit( X_scaled )
 cluster_errors.append( clusters.inertia_ )
 clusters_df = pd.DataFrame( { "num_clusters":cluster_range, "cluster_errors": cluster_errors } )
#clusters_df[0:10]
#
#plt.figure(figsize=(10,7))
#plt.plot( clusters_df.num_clusters, clusters_df.cluster_errors, marker = "o" )
#plt.xlabel('Number of clusters')
#plt.ylabel('WSS value')
```



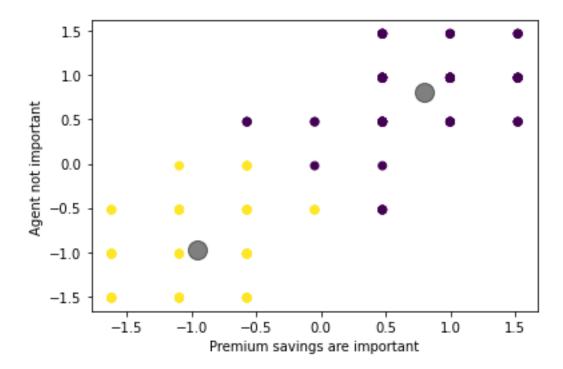
#the 'elbow' is at k == 2

```
kmeans = KMeans(n_clusters=2)
kmeans.fit(X_scaled)
y_kmeans = kmeans.predict(X_scaled)
```

#y\_kmeans

```
centers = kmeans.cluster_centers_
#centers
```

```
plt.scatter(X_scaled[:, 0], X_scaled[:, 1], c=y_kmeans)
plt.scatter(centers[:, 0], centers[:, 1], c='black', s=200, alpha=0.5)
plt.xlabel('Premium savings are important')
plt.ylabel('Agent not important')
```



#Comparing it with k = 3

#kmeans = KMeans(n\_clusters=3)

#kmeans.fit(X\_scaled)

#y\_kmeans = kmeans.predict(X\_scaled)

#y\_kmeans

#centers = kmeans.cluster\_centers\_

#centers

#plt.scatter(X\_scaled[:, 0], X\_scaled[:, 1], c=y\_kmeans)

#plt.scatter(centers[:, 0], centers[:, 1], c='black', s=200, alpha=0.5)

#plt.xlabel('Premium savings are important')

#plt.ylabel('Agent not important')

