

Overview

Customer Analysis

Customer Segmentation

# Customer Segmentation

After evaluating various models( kmeans, ap, meanshift, dbscan, kmodes, etc.), the best silhouette score is shown in kmeans. With high dimension data as ours, we use PCA to reduced dimension of data before clustering process using kmeans.

## K-Mean with PCA Clustering

Original Data (21 dimensions)

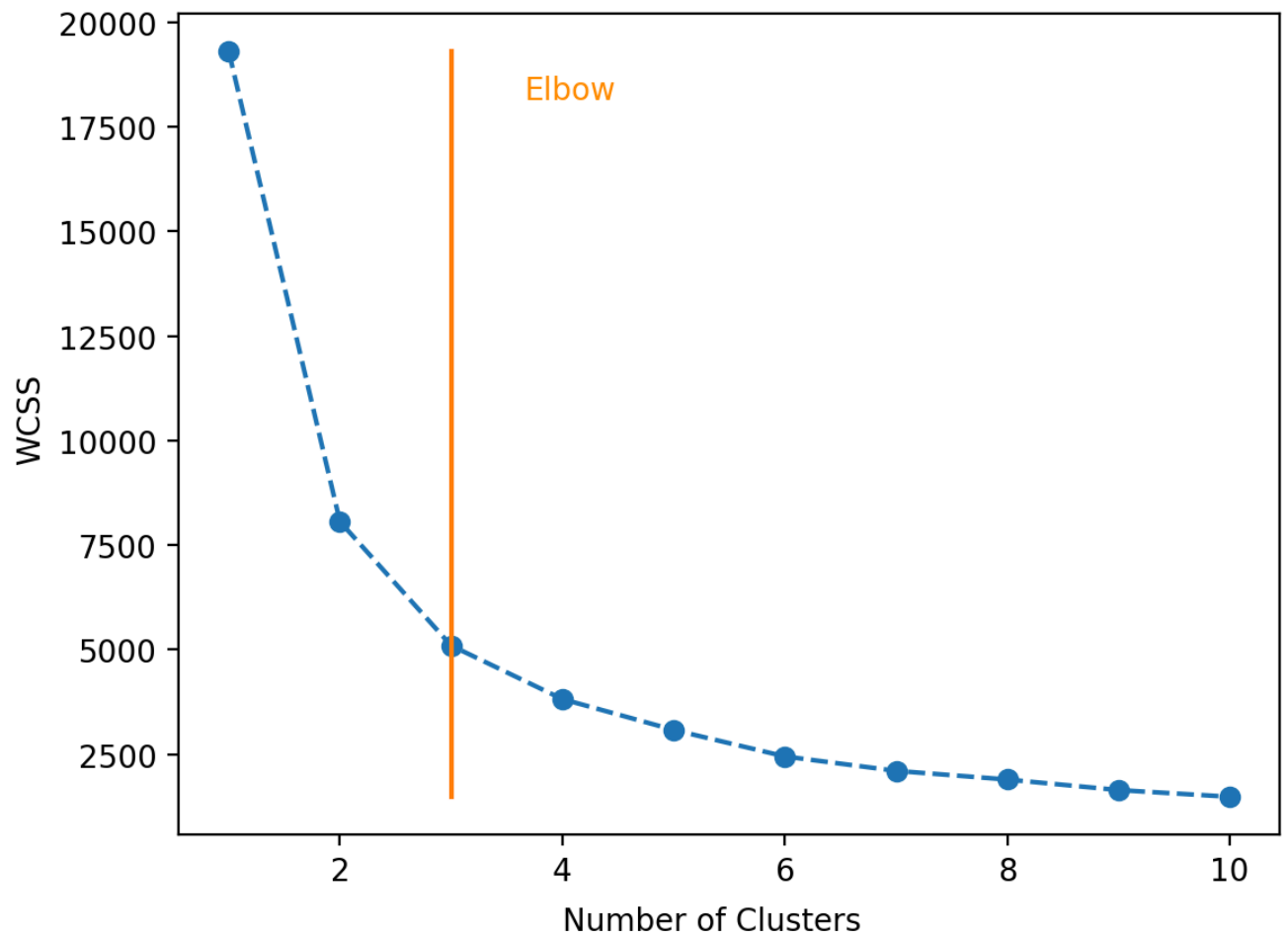
	Income	Kidhome	Teenhome	Recency	MntWines	MntFruits
0	58,138	0	0	58	635	88
1	46,344	1	1	38	11	1
2	71,613	0	0	26	426	49
3	26,646	1	0	26	11	4
4	58,293	1	0	94	173	43
5	62,513	0	1	16	520	42
6	55,635	0	1	34	235	65
7	33,454	1	0	32	76	10
8	30,351	1	0	19	14	0
9	5,648	1	1	68	28	0

Reduced-dimension (2 PCs)

	PC1	PC2
0	3.9061	-1.3603
1	-2.2067	-0.6226
2	1.9702	-0.7049
3	-2.5575	-1.6737
4	-0.2907	0.9669
5	0.9335	2.3464
6	0.9001	-0.4491
7	-2.3709	0.6265
8	-2.9557	0.681
9	-4.8077	1.9284

## WCSS vs Number of Clusters

The “elbow” point indicates a strong candidate for the optimal number of clusters at 3.



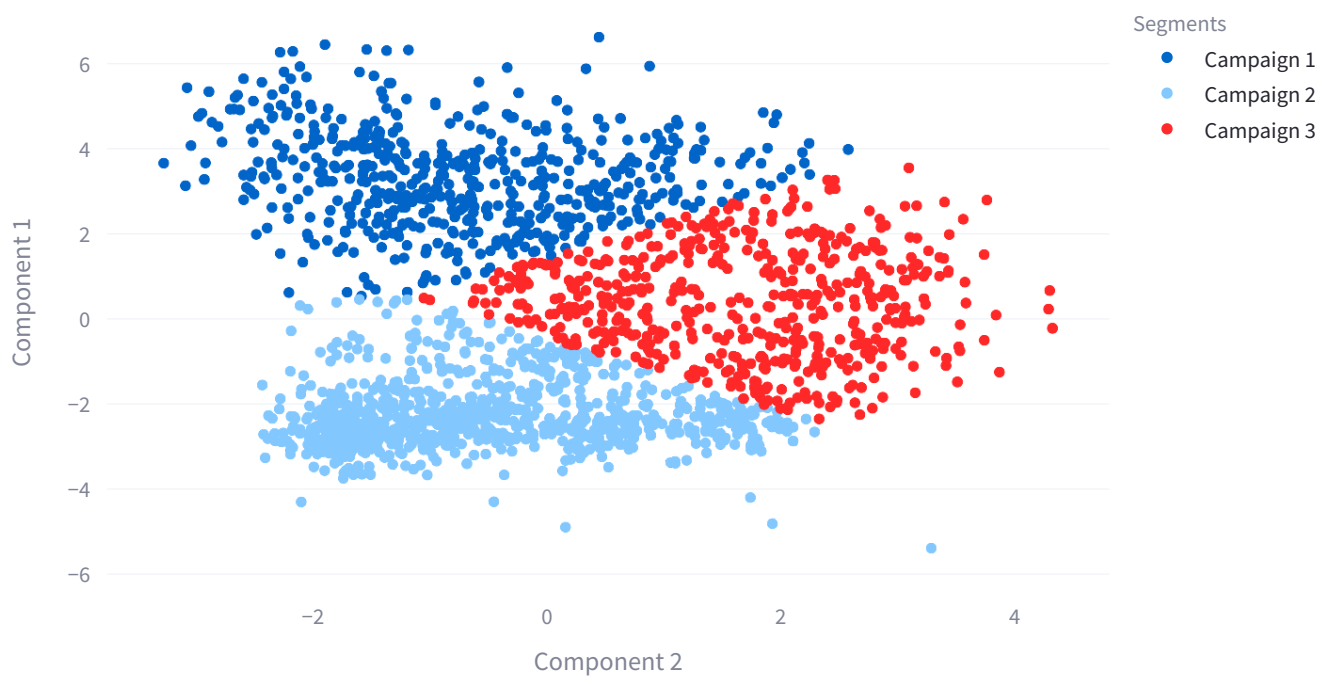
## Customer Segments

Merge the 2 PCA components with original data. Apply number of Cluster to generate segments.

WCSS shows elbow at K=3



	Income (\$)	Kidhome	Teenhome	Recency	MntWines	MntFruits	MntMeatProducts	MntFishProducts
0	58,138	0	0	58	635	88	546	17
1	46,344	1	1	38	11	1	6	
2	71,613	0	0	26	426	49	127	10
3	26,646	1	0	26	11	4	20	5
4	58,293	1	0	94	173	43	118	4
5	62,513	0	1	16	520	42	98	
6	55,635	0	1	34	235	65	164	5
7	33,454	1	0	32	76	10	56	
8	30,351	1	0	19	14	0	24	
9	5,648	1	1	68	28	0	6	



Each segmentation contains customer with different behavior. We can drive further to find the significant factors in which campaign can be launched based upon.

