

## 22BCE0476AMAN.R

amssr

2024-10-03

```
#CustomerID Genre Age Annual Income (k$) Spending Score (1-100)
# K-Means Clustering

# Importing the dataset - Mall_Customers and extract 4th and 5th column

# Splitting the dataset into the Training set and Test set - Load the Library
caTools

#Identify the Dependent variable and split the data into 80% for training and
20% for testing

# Feature Scaling
# Using the elbow method to find the optimal number of clusters - wcss
# Fitting K-Means to the dataset

#Identify the the number of cluster and plot

#Mall_Customers.csv Mall_Customers.csv

#install.packages("caTools")
library(caTools)

## Warning: package 'caTools' was built under R version 4.3.3

#install.packages("cluster")
library(cluster)

## Warning: package 'cluster' was built under R version 4.3.3

dataset <- read.csv("C:/Users/amssr/Desktop/Mall_Customers.csv")
dataset
```

	CustomerID	Genre	Age	Annual.Income..k..	Spending.Score..1.100.
## 1	1	Male	19	15	39
## 2	2	Male	21	15	81
## 3	3	Female	20	16	6
## 4	4	Female	23	16	77
## 5	5	Female	31	17	40
## 6	6	Female	22	17	76
## 7	7	Female	35	18	6
## 8	8	Female	23	18	94
## 9	9	Male	64	19	3

## 10	10 Female	30	19	72
## 11	11 Male	67	19	14
## 12	12 Female	35	19	99
## 13	13 Female	58	20	15
## 14	14 Female	24	20	77
## 15	15 Male	37	20	13
## 16	16 Male	22	20	79
## 17	17 Female	35	21	35
## 18	18 Male	20	21	66
## 19	19 Male	52	23	29
## 20	20 Female	35	23	98
## 21	21 Male	35	24	35
## 22	22 Male	25	24	73
## 23	23 Female	46	25	5
## 24	24 Male	31	25	73
## 25	25 Female	54	28	14
## 26	26 Male	29	28	82
## 27	27 Female	45	28	32
## 28	28 Male	35	28	61
## 29	29 Female	40	29	31
## 30	30 Female	23	29	87
## 31	31 Male	60	30	4
## 32	32 Female	21	30	73
## 33	33 Male	53	33	4
## 34	34 Male	18	33	92
## 35	35 Female	49	33	14
## 36	36 Female	21	33	81
## 37	37 Female	42	34	17
## 38	38 Female	30	34	73
## 39	39 Female	36	37	26
## 40	40 Female	20	37	75
## 41	41 Female	65	38	35
## 42	42 Male	24	38	92
## 43	43 Male	48	39	36
## 44	44 Female	31	39	61
## 45	45 Female	49	39	28
## 46	46 Female	24	39	65
## 47	47 Female	50	40	55
## 48	48 Female	27	40	47
## 49	49 Female	29	40	42
## 50	50 Female	31	40	42
## 51	51 Female	49	42	52
## 52	52 Male	33	42	60
## 53	53 Female	31	43	54
## 54	54 Male	59	43	60
## 55	55 Female	50	43	45
## 56	56 Male	47	43	41
## 57	57 Female	51	44	50
## 58	58 Male	69	44	46
## 59	59 Female	27	46	51

## 60	60	Male	53	46	46
## 61	61	Male	70	46	56
## 62	62	Male	19	46	55
## 63	63	Female	67	47	52
## 64	64	Female	54	47	59
## 65	65	Male	63	48	51
## 66	66	Male	18	48	59
## 67	67	Female	43	48	50
## 68	68	Female	68	48	48
## 69	69	Male	19	48	59
## 70	70	Female	32	48	47
## 71	71	Male	70	49	55
## 72	72	Female	47	49	42
## 73	73	Female	60	50	49
## 74	74	Female	60	50	56
## 75	75	Male	59	54	47
## 76	76	Male	26	54	54
## 77	77	Female	45	54	53
## 78	78	Male	40	54	48
## 79	79	Female	23	54	52
## 80	80	Female	49	54	42
## 81	81	Male	57	54	51
## 82	82	Male	38	54	55
## 83	83	Male	67	54	41
## 84	84	Female	46	54	44
## 85	85	Female	21	54	57
## 86	86	Male	48	54	46
## 87	87	Female	55	57	58
## 88	88	Female	22	57	55
## 89	89	Female	34	58	60
## 90	90	Female	50	58	46
## 91	91	Female	68	59	55
## 92	92	Male	18	59	41
## 93	93	Male	48	60	49
## 94	94	Female	40	60	40
## 95	95	Female	32	60	42
## 96	96	Male	24	60	52
## 97	97	Female	47	60	47
## 98	98	Female	27	60	50
## 99	99	Male	48	61	42
## 100	100	Male	20	61	49
## 101	101	Female	23	62	41
## 102	102	Female	49	62	48
## 103	103	Male	67	62	59
## 104	104	Male	26	62	55
## 105	105	Male	49	62	56
## 106	106	Female	21	62	42
## 107	107	Female	66	63	50
## 108	108	Male	54	63	46
## 109	109	Male	68	63	43

## 110	110	Male	66	63	48
## 111	111	Male	65	63	52
## 112	112	Female	19	63	54
## 113	113	Female	38	64	42
## 114	114	Male	19	64	46
## 115	115	Female	18	65	48
## 116	116	Female	19	65	50
## 117	117	Female	63	65	43
## 118	118	Female	49	65	59
## 119	119	Female	51	67	43
## 120	120	Female	50	67	57
## 121	121	Male	27	67	56
## 122	122	Female	38	67	40
## 123	123	Female	40	69	58
## 124	124	Male	39	69	91
## 125	125	Female	23	70	29
## 126	126	Female	31	70	77
## 127	127	Male	43	71	35
## 128	128	Male	40	71	95
## 129	129	Male	59	71	11
## 130	130	Male	38	71	75
## 131	131	Male	47	71	9
## 132	132	Male	39	71	75
## 133	133	Female	25	72	34
## 134	134	Female	31	72	71
## 135	135	Male	20	73	5
## 136	136	Female	29	73	88
## 137	137	Female	44	73	7
## 138	138	Male	32	73	73
## 139	139	Male	19	74	10
## 140	140	Female	35	74	72
## 141	141	Female	57	75	5
## 142	142	Male	32	75	93
## 143	143	Female	28	76	40
## 144	144	Female	32	76	87
## 145	145	Male	25	77	12
## 146	146	Male	28	77	97
## 147	147	Male	48	77	36
## 148	148	Female	32	77	74
## 149	149	Female	34	78	22
## 150	150	Male	34	78	90
## 151	151	Male	43	78	17
## 152	152	Male	39	78	88
## 153	153	Female	44	78	20
## 154	154	Female	38	78	76
## 155	155	Female	47	78	16
## 156	156	Female	27	78	89
## 157	157	Male	37	78	1
## 158	158	Female	30	78	78
## 159	159	Male	34	78	1

## 160	160	Female	30	78	73
## 161	161	Female	56	79	35
## 162	162	Female	29	79	83
## 163	163	Male	19	81	5
## 164	164	Female	31	81	93
## 165	165	Male	50	85	26
## 166	166	Female	36	85	75
## 167	167	Male	42	86	20
## 168	168	Female	33	86	95
## 169	169	Female	36	87	27
## 170	170	Male	32	87	63
## 171	171	Male	40	87	13
## 172	172	Male	28	87	75
## 173	173	Male	36	87	10
## 174	174	Male	36	87	92
## 175	175	Female	52	88	13
## 176	176	Female	30	88	86
## 177	177	Male	58	88	15
## 178	178	Male	27	88	69
## 179	179	Male	59	93	14
## 180	180	Male	35	93	90
## 181	181	Female	37	97	32
## 182	182	Female	32	97	86
## 183	183	Male	46	98	15
## 184	184	Female	29	98	88
## 185	185	Female	41	99	39
## 186	186	Male	30	99	97
## 187	187	Female	54	101	24
## 188	188	Male	28	101	68
## 189	189	Female	41	103	17
## 190	190	Female	36	103	85
## 191	191	Female	34	103	23
## 192	192	Female	32	103	69
## 193	193	Male	33	113	8
## 194	194	Female	38	113	91
## 195	195	Female	47	120	16
## 196	196	Female	35	120	79
## 197	197	Female	45	126	28
## 198	198	Male	32	126	74
## 199	199	Male	32	137	18
## 200	200	Male	30	137	83

```
X <- dataset[, 4:5]
```

```
set.seed(123)
```

```
split <- sample.split(X, SplitRatio = 0.8)
training_set <- subset(X, split == TRUE)
test_set <- subset(X, split == FALSE)
```

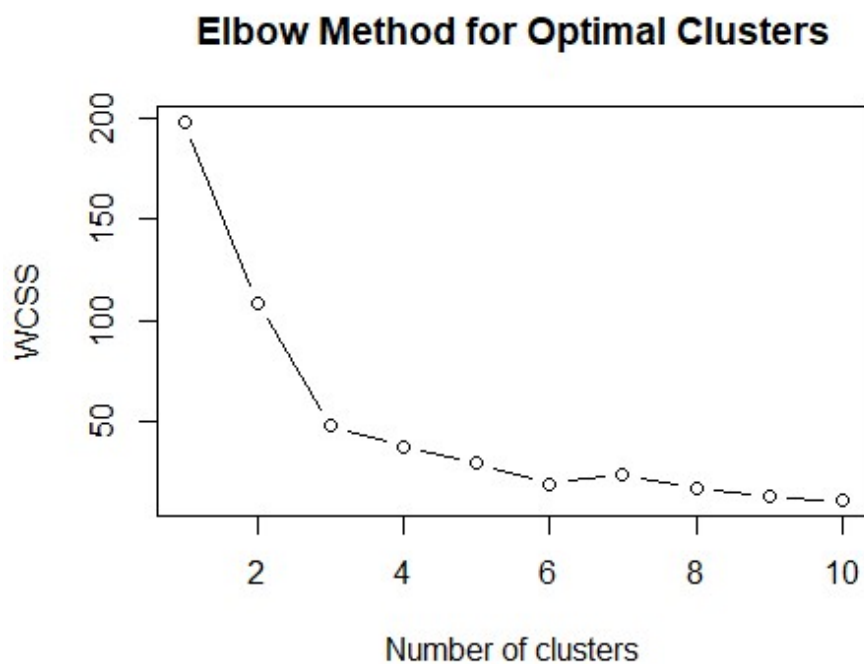
```

training_set_scaled <- scale(training_set)
test_set_scaled <- scale(test_set)

# Using the elbow method to find the optimal number of clusters
set.seed(123)
wcss <- vector()
for (i in 1:10) {
  kmeans_model <- kmeans(training_set_scaled, centers = i)
  wcss[i] <- kmeans_model$tot.withinss
}

# Plot the elbow graph
plot(1:10, wcss, type = "b", main = "Elbow Method for Optimal Clusters", xlab = "Number of clusters", ylab = "WCSS")

```



```

#5. Fitting K-Means to the Dataset
# Fitting K-Means to the dataset
set.seed(123)
kmeans_model <- kmeans(training_set_scaled, centers = 5, nstart = 25)

#visualize the clusters

library(cluster)

```

```
clusplot(training_set_scaled,
         kmeans_model$cluster,
         lines = 0,
         shade = TRUE,
         color = TRUE,
         labels = 2,
         plotchar = FALSE,
         span = TRUE,
         main = "Clusters of customers",
         xlab = "Annual Income (k$)",
         ylab = "Spending Score (1-100)")
```

