

Assignment2_DecisionTree

Mukhtar

3/28/2022

```
library(ISLR)
library(rpart)
library(rattle)
```

```
## Warning: package 'rattle' was built under R version 4.1.2
```

```
## Loading required package: tibble
```

```
## Loading required package: bitops
```

```
## Rattle: A free graphical interface for data science with R.
## Version 5.4.0 Copyright (c) 2006-2020 Togaware Pty Ltd.
## Type 'rattle()' to shake, rattle, and roll your data.
```

```
library(rpart.plot)
```

```
## Warning: package 'rpart.plot' was built under R version 4.1.2
```

```
library(caret)
```

```
## Loading required package: ggplot2
```

```
## Loading required package: lattice
```

```
library(ggplot2)
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
##
##   filter, lag
```

```
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

R Markdown

This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more details on using R Markdown see <http://rmarkdown.rstudio.com>.

Problem Statement

Data Preparation

```
getwd()
```

```
## [1] "C:/Users/Mukht/OneDrive/Desktop/Kent State University/54050-Project-ADM/Assignment2"
```

```
setwd("C:\\Users\\Mukht\\OneDrive\\Desktop\\Kent State University\\54050-Project-ADM\\Assignment2")
```

```
ADM_Assignment2<-read.csv("carseats_ADM.csv")
str(ADM_Assignment2)
```

```
## 'data.frame': 400 obs. of 11 variables:
## $ i..Sales : num 9.5 11.22 10.06 7.4 4.15 ...
## $ Income : int 73 48 35 100 64 113 105 81 110 113 ...
## $ Advertising: int 11 16 10 4 3 13 0 15 0 0 ...
## $ Population : int 276 260 269 466 340 501 45 425 108 131 ...
## $ Price : int 120 83 80 97 128 72 108 120 124 124 ...
## $ Age : int 42 65 59 55 38 78 71 67 76 76 ...
## $ Education : int 17 10 12 14 13 16 15 10 10 17 ...
## $ Urban : chr "Yes" "Yes" "Yes" "Yes" ...
## $ US : chr "Yes" "Yes" "Yes" "Yes" ...
## $ CompPrice : int 138 111 113 117 141 124 115 136 132 132 ...
## $ ShelveLoc : chr "Bad" "Good" "Medium" "Medium" ...
```

```
head(ADM_Assignment2)
```

```
## i..Sales Income Advertising Population Price Age Education Urban US
## 1 9.50 73 11 276 120 42 17 Yes Yes
## 2 11.22 48 16 260 83 65 10 Yes Yes
## 3 10.06 35 10 269 80 59 12 Yes Yes
## 4 7.40 100 4 466 97 55 14 Yes Yes
## 5 4.15 64 3 340 128 38 13 Yes No
## 6 10.81 113 13 501 72 78 16 No Yes
## CompPrice ShelveLoc
## 1 138 Bad
## 2 111 Good
## 3 113 Medium
## 4 117 Medium
## 5 141 Bad
## 6 124 Bad
```

#Three of the variables are factors, while the rest are numeric. Currently there are no missing observations.

```
Carseats_Filtered <- ADM_Assignment2 %>% select("i..Sales", "Price", "Advertising", "Population", "Age", "Income", "Education")
Carseats_Filtered
```

##	i..Sales	Price	Advertising	Population	Age	Income	Education
## 1	9.50	120	11	276	42	73	17
## 2	11.22	83	16	260	65	48	10
## 3	10.06	80	10	269	59	35	12
## 4	7.40	97	4	466	55	100	14
## 5	4.15	128	3	340	38	64	13
## 6	10.81	72	13	501	78	113	16
## 7	6.63	108	0	45	71	105	15
## 8	11.85	120	15	425	67	81	10
## 9	6.54	124	0	108	76	110	10
## 10	4.69	124	0	131	76	113	17
## 11	9.01	100	9	150	26	78	10
## 12	11.96	94	4	503	50	94	13
## 13	3.98	136	2	393	62	35	18
## 14	10.96	86	11	29	53	28	18
## 15	11.17	118	11	148	52	117	18
## 16	8.71	144	5	400	76	95	18
## 17	7.58	110	0	284	63	32	13
## 18	12.29	131	13	251	52	74	10
## 19	13.91	68	0	408	46	110	17
## 20	8.73	121	16	58	69	76	12
## 21	6.41	131	2	367	35	90	18
## 22	12.13	109	12	239	62	29	18
## 23	5.08	138	6	497	42	46	13
## 24	5.87	109	0	292	79	31	10
## 25	10.14	113	16	294	42	119	12
## 26	14.90	82	0	176	54	32	11
## 27	8.33	131	11	496	50	115	11
## 28	5.27	107	0	19	64	118	17
## 29	2.99	97	0	359	55	74	11
## 30	7.81	102	15	226	58	99	17
## 31	13.55	89	0	447	30	94	12
## 32	8.25	131	16	241	44	58	18
## 33	6.20	137	12	236	64	32	10
## 34	8.77	128	13	317	50	38	16
## 35	2.67	128	0	406	42	54	17
## 36	11.07	96	11	29	44	84	17
## 37	8.89	100	0	270	60	76	18
## 38	4.95	110	5	412	54	41	10
## 39	6.59	102	0	454	65	73	15
## 40	3.24	138	0	144	38	60	10
## 41	2.07	126	0	18	73	98	17
## 42	7.96	124	0	403	58	53	16
## 43	10.43	24	0	25	50	69	18
## 44	4.12	134	11	16	59	42	13
## 45	4.16	95	6	325	69	79	13
## 46	4.56	135	0	168	44	63	12
## 47	12.44	70	14	16	48	90	15
## 48	4.38	108	0	173	55	98	16

## 49	3.91	98	0	349	69	52	18
## 50	10.61	149	0	51	32	93	17
## 51	1.42	108	18	341	80	32	16
## 52	4.42	108	0	150	75	90	16
## 53	7.91	129	3	112	39	40	18
## 54	6.92	119	13	39	61	64	17
## 55	4.90	144	13	25	76	103	17
## 56	6.85	154	5	60	61	81	18
## 57	11.91	84	0	54	50	82	17
## 58	0.91	117	0	22	75	91	11
## 59	5.42	103	15	188	74	93	16
## 60	5.21	114	4	148	80	71	13
## 61	8.32	123	19	469	29	102	13
## 62	7.32	107	0	358	26	32	13
## 63	1.82	133	0	146	77	45	17
## 64	8.47	101	10	170	61	88	13
## 65	7.80	104	12	184	32	67	16
## 66	4.90	128	0	197	55	26	13
## 67	8.85	91	0	508	56	92	18
## 68	9.01	115	14	152	47	61	16
## 69	13.39	134	20	366	60	69	13
## 70	7.99	99	0	339	65	59	12
## 71	9.46	99	15	237	74	81	12
## 72	6.50	150	16	148	58	51	17
## 73	5.52	116	0	432	25	45	15
## 74	12.61	104	10	54	31	90	11
## 75	6.20	136	5	125	64	68	13
## 76	8.55	92	23	480	36	111	16
## 77	10.64	70	10	346	64	87	15
## 78	7.70	89	12	44	67	71	18
## 79	4.43	145	1	139	65	48	12
## 80	9.14	90	0	286	41	67	13
## 81	8.01	79	16	353	68	100	11
## 82	7.52	128	0	237	70	72	13
## 83	11.62	139	4	325	28	83	17
## 84	4.42	94	7	468	56	36	11
## 85	2.23	121	0	52	43	25	18
## 86	8.47	112	0	304	49	103	13
## 87	8.70	134	9	432	64	84	15
## 88	11.70	126	7	272	54	67	16
## 89	6.56	111	7	144	62	42	10
## 90	7.95	119	3	493	45	66	16
## 91	5.33	103	0	491	64	22	11
## 92	4.81	107	11	267	80	46	15
## 93	4.53	125	0	97	29	113	12
## 94	8.86	104	0	67	55	30	17
## 95	8.39	84	5	134	55	97	11
## 96	5.58	148	10	237	59	25	13
## 97	9.48	132	10	407	73	42	16
## 98	7.45	129	5	287	33	82	16
## 99	12.49	127	24	382	36	77	16
## 100	4.88	107	3	220	56	47	16
## 101	4.11	106	11	94	76	69	12
## 102	6.20	118	0	89	34	93	18

## 103	5.30	97	0	57	65	22	16
## 104	5.07	96	0	334	78	91	17
## 105	4.62	138	0	472	51	96	12
## 106	5.55	97	8	398	61	100	11
## 107	0.16	139	0	217	70	33	18
## 108	8.55	108	0	104	60	107	12
## 109	3.47	103	2	488	65	79	16
## 110	8.98	90	0	217	60	65	17
## 111	9.00	116	7	125	43	62	14
## 112	6.62	151	12	272	43	118	14
## 113	6.67	125	5	298	62	99	12
## 114	6.01	127	11	335	33	29	12
## 115	9.31	106	9	17	65	87	13
## 116	8.54	129	0	95	42	35	13
## 117	5.08	128	0	202	80	75	10
## 118	8.80	119	0	507	41	53	12
## 119	7.57	99	2	243	62	88	11
## 120	7.37	128	8	137	64	94	12
## 121	6.87	131	11	249	63	105	13
## 122	11.67	87	10	380	28	89	10
## 123	6.88	108	5	45	75	100	10
## 124	8.19	155	0	125	29	103	15
## 125	8.87	120	0	181	63	113	14
## 126	9.34	49	0	181	43	78	15
## 127	11.27	133	2	60	59	68	16
## 128	6.52	116	3	192	51	48	14
## 129	4.96	126	3	350	55	100	13
## 130	4.47	147	7	279	40	120	10
## 131	8.41	77	13	497	51	84	12
## 132	6.50	94	3	208	77	69	16
## 133	9.54	136	9	232	72	87	10
## 134	7.62	97	2	265	62	98	12
## 135	3.67	131	0	327	76	31	16
## 136	6.44	120	14	384	36	94	18
## 137	5.17	120	0	10	31	75	18
## 138	6.52	118	0	436	80	42	11
## 139	10.27	109	12	371	44	103	10
## 140	12.30	94	10	310	30	62	13
## 141	6.03	129	10	277	45	60	18
## 142	6.53	131	0	331	28	42	15
## 143	7.44	104	0	300	77	84	15
## 144	0.53	159	7	36	28	88	17
## 145	9.09	123	0	264	34	68	11
## 146	8.77	117	11	27	47	63	17
## 147	3.90	131	0	412	39	83	14
## 148	10.51	119	9	402	41	54	16
## 149	7.56	97	0	384	72	119	14
## 150	11.48	87	13	140	56	120	11
## 151	10.49	114	8	176	57	84	10
## 152	10.77	103	17	407	75	58	17
## 153	7.64	128	0	341	45	78	13
## 154	5.93	150	7	488	25	36	17
## 155	6.89	110	10	289	50	69	16
## 156	7.71	69	0	59	65	72	16

## 157	7.49	157	0	220	51	34	16
## 158	10.21	90	8	249	48	58	13
## 159	12.53	112	1	189	39	90	10
## 160	9.32	70	0	372	30	60	18
## 161	4.67	111	0	486	29	28	12
## 162	2.93	160	5	81	67	21	12
## 163	3.63	149	0	424	51	74	13
## 164	5.68	106	0	40	39	64	17
## 165	8.22	141	0	58	27	64	13
## 166	0.37	191	7	100	27	58	15
## 167	6.71	137	17	151	55	67	11
## 168	6.71	93	0	216	60	73	13
## 169	7.30	117	0	425	45	89	10
## 170	11.48	77	15	492	73	41	18
## 171	8.01	118	12	356	71	39	10
## 172	12.49	55	12	416	75	106	15
## 173	9.03	110	13	123	35	102	16
## 174	6.38	128	5	207	66	91	18
## 175	0.00	185	0	358	79	24	15
## 176	7.54	122	0	38	25	89	12
## 177	5.61	154	9	480	47	107	11
## 178	10.48	94	0	148	27	72	17
## 179	10.66	81	14	89	25	71	14
## 180	7.78	116	3	70	77	25	18
## 181	4.94	149	15	434	66	112	13
## 182	7.43	91	0	79	68	83	11
## 183	4.74	140	4	230	25	60	13
## 184	5.32	102	6	426	80	74	18
## 185	9.95	97	7	35	60	33	11
## 186	10.07	107	11	449	64	100	10
## 187	8.68	86	0	93	46	51	17
## 188	6.03	96	0	142	62	32	17
## 189	8.07	90	0	426	76	37	15
## 190	12.11	104	18	509	26	117	15
## 191	8.79	101	13	297	37	37	13
## 192	6.67	173	13	170	74	42	14
## 193	7.56	93	0	408	56	26	14
## 194	13.28	96	7	71	61	70	10
## 195	7.23	128	18	481	45	98	11
## 196	4.19	112	4	420	66	93	11
## 197	4.10	133	6	410	72	28	16
## 198	2.52	138	0	333	76	61	16
## 199	3.62	128	5	500	69	80	10
## 200	6.42	126	5	335	64	88	14
## 201	5.56	146	0	349	62	92	12
## 202	5.94	134	0	139	54	83	18
## 203	4.10	130	4	413	46	78	10
## 204	2.05	157	0	132	25	82	14
## 205	8.74	124	0	237	37	80	14
## 206	5.68	132	1	317	28	22	12
## 207	4.97	160	0	27	77	67	17
## 208	8.19	97	0	466	61	105	10
## 209	7.78	64	0	497	33	54	12
## 210	3.02	90	11	326	76	21	11

## 211	4.36	123	2	357	47	41	14
## 212	9.39	120	14	445	32	118	15
## 213	12.04	105	19	501	45	69	11
## 214	8.23	139	5	220	33	84	10
## 215	4.83	107	3	48	73	115	18
## 216	2.34	144	15	170	71	83	11
## 217	5.73	144	0	243	34	33	17
## 218	4.34	111	0	481	70	44	14
## 219	9.70	120	12	156	25	61	14
## 220	10.62	116	19	359	58	79	17
## 221	10.59	124	15	262	30	120	10
## 222	6.43	107	0	125	80	44	11
## 223	7.49	145	6	178	35	119	13
## 224	3.45	125	9	276	62	45	14
## 225	4.10	141	0	464	48	82	13
## 226	6.68	82	0	412	36	25	14
## 227	7.80	122	0	245	56	33	14
## 228	8.69	101	10	68	57	64	16
## 229	5.40	163	13	381	26	73	11
## 230	11.19	72	0	404	27	104	18
## 231	5.16	114	0	119	38	60	14
## 232	8.09	122	0	123	27	69	11
## 233	13.14	105	10	24	61	80	15
## 234	8.65	120	18	218	29	76	14
## 235	9.43	129	11	289	56	62	16
## 236	5.53	132	8	95	50	32	17
## 237	9.32	108	16	361	69	34	10
## 238	9.62	135	8	499	48	28	10
## 239	7.36	133	0	200	73	24	13
## 240	3.89	118	0	149	62	105	16
## 241	10.31	121	0	362	26	80	18
## 242	12.01	94	0	160	38	63	12
## 243	4.68	135	0	199	52	46	14
## 244	7.82	110	13	87	57	25	10
## 245	8.78	100	0	391	26	30	18
## 246	10.00	88	0	199	57	43	10
## 247	6.90	90	20	266	78	56	18
## 248	5.04	151	0	298	34	114	16
## 249	5.36	101	0	12	61	52	11
## 250	5.05	117	0	86	65	67	11
## 251	9.16	156	10	435	72	105	14
## 252	3.72	132	5	310	62	111	13
## 253	8.31	117	0	70	32	97	16
## 254	5.64	122	5	288	57	24	12
## 255	9.58	129	23	353	37	104	17
## 256	7.71	81	8	198	80	81	15
## 257	4.20	144	0	277	73	40	10
## 258	8.67	112	14	477	80	62	13
## 259	3.47	81	0	251	72	38	14
## 260	5.12	100	10	467	74	36	11
## 261	7.67	101	8	400	36	117	10
## 262	5.71	118	4	188	54	42	15
## 263	6.37	132	15	86	48	77	18
## 264	7.77	115	6	434	25	26	17

## 265	6.95	159	5	324	31	29	15
## 266	5.31	129	10	402	39	35	17
## 267	9.10	112	12	343	73	93	17
## 268	5.83	112	7	473	51	82	12
## 269	6.53	105	0	66	39	57	11
## 270	5.01	166	0	438	46	69	17
## 271	11.99	89	0	284	26	26	10
## 272	4.55	110	0	504	62	56	16
## 273	12.98	63	0	14	38	33	12
## 274	10.04	86	8	244	58	106	12
## 275	7.22	119	2	67	34	93	11
## 276	6.67	132	11	210	53	119	11
## 277	6.93	130	14	296	73	69	15
## 278	7.80	125	12	326	36	48	16
## 279	7.22	151	2	129	40	113	15
## 280	3.42	158	13	376	64	57	18
## 281	2.86	145	10	496	51	86	10
## 282	11.19	105	7	303	45	69	16
## 283	7.74	154	0	80	61	96	11
## 284	5.36	117	0	112	80	110	16
## 285	6.97	96	11	414	79	46	17
## 286	7.60	131	11	261	39	26	10
## 287	7.53	113	11	429	67	118	18
## 288	6.88	72	4	208	44	44	17
## 289	6.98	97	0	74	76	40	15
## 290	8.75	156	25	448	43	77	17
## 291	9.49	103	14	400	41	111	11
## 292	6.64	89	0	106	39	70	17
## 293	11.82	74	16	322	76	66	15
## 294	11.28	89	0	74	59	84	10
## 295	12.66	99	3	126	60	76	11
## 296	4.21	137	14	502	79	35	10
## 297	8.21	123	13	160	63	44	18
## 298	3.07	104	13	276	75	83	10
## 299	10.98	130	0	312	63	63	15
## 300	9.40	96	17	497	54	40	17
## 301	8.57	99	1	158	45	78	11
## 302	7.41	87	0	198	57	93	16
## 303	5.28	110	13	388	74	77	14
## 304	10.01	99	16	290	43	52	11
## 305	11.93	134	12	408	29	98	10
## 306	8.03	132	26	394	33	29	13
## 307	4.78	133	1	85	48	32	12
## 308	5.90	120	0	13	61	92	12
## 309	9.24	126	19	436	52	80	10
## 310	11.18	80	13	33	68	111	18
## 311	9.53	166	29	419	53	65	12
## 312	6.15	132	12	328	51	68	14
## 313	6.80	135	5	337	38	117	10
## 314	9.33	54	3	491	66	81	13
## 315	7.72	129	10	333	71	33	14
## 316	6.39	171	8	220	29	21	14
## 317	15.63	72	5	369	35	36	10
## 318	6.41	136	0	472	80	30	15

## 319	10.08	130	10	456	41	72	14
## 320	6.97	129	19	459	57	45	11
## 321	5.86	152	12	171	44	70	18
## 322	7.52	98	5	499	34	39	15
## 323	9.16	139	10	300	60	50	15
## 324	10.36	103	18	428	34	105	12
## 325	2.66	150	4	133	53	65	13
## 326	11.70	104	11	131	47	69	11
## 327	4.69	122	0	152	53	30	17
## 328	6.23	104	17	316	80	38	16
## 329	3.15	111	1	65	55	66	11
## 330	11.27	89	9	433	45	54	12
## 331	4.99	112	0	501	32	59	14
## 332	10.10	134	15	213	32	63	10
## 333	5.74	104	20	354	61	33	12
## 334	5.87	147	7	303	41	60	10
## 335	7.63	83	9	489	42	117	13
## 336	6.18	110	15	464	72	70	15
## 337	5.17	143	6	60	28	35	18
## 338	8.61	102	0	283	80	38	15
## 339	5.97	101	0	164	45	24	11
## 340	11.54	126	4	219	44	44	15
## 341	7.50	91	0	105	43	29	16
## 342	7.38	93	0	268	72	120	10
## 343	7.81	118	13	422	71	102	10
## 344	5.99	121	10	371	26	42	14
## 345	8.43	126	0	108	70	80	13
## 346	4.81	149	0	279	79	68	12
## 347	8.97	125	0	144	33	107	13
## 348	6.88	112	0	161	27	39	14
## 349	12.57	107	20	459	49	102	11
## 350	9.32	96	18	467	49	27	14
## 351	8.64	91	17	266	63	101	17
## 352	10.44	105	16	458	62	115	16
## 353	13.44	122	14	288	61	103	17
## 354	9.45	92	12	430	35	67	12
## 355	5.30	145	1	80	42	31	18
## 356	7.02	146	0	306	42	100	11
## 357	3.58	164	0	111	72	109	12
## 358	13.36	72	3	276	34	73	15
## 359	4.17	118	10	71	69	96	11
## 360	3.13	130	11	396	66	62	14
## 361	8.77	114	7	265	52	86	15
## 362	8.68	104	10	183	56	25	15
## 363	5.25	110	0	26	79	55	12
## 364	10.26	108	1	377	25	75	12
## 365	10.50	131	16	488	30	21	14
## 366	6.53	162	0	122	57	30	17
## 367	5.98	134	11	447	53	56	12
## 368	14.37	53	0	256	52	106	17
## 369	10.71	79	10	348	74	22	14
## 370	10.26	122	22	463	36	100	14
## 371	7.68	119	22	403	42	41	12
## 372	9.08	126	0	191	54	81	16

## 373	7.80	98	0	508	65	50	11
## 374	5.58	116	0	402	78	71	17
## 375	9.44	118	7	90	47	47	12
## 376	7.90	124	4	206	73	46	11
## 377	16.27	92	19	319	44	60	11
## 378	6.81	125	0	263	41	61	12
## 379	6.11	119	3	105	79	88	12
## 380	5.81	107	0	404	54	111	15
## 381	9.64	89	10	17	68	64	17
## 382	3.90	151	21	496	77	65	13
## 383	4.95	121	19	315	66	28	14
## 384	9.35	68	0	76	63	117	10
## 385	12.85	112	15	348	28	37	12
## 386	5.87	132	13	455	62	73	17
## 387	5.32	160	0	170	39	116	16
## 388	8.67	115	14	238	73	73	14
## 389	8.14	78	11	245	79	89	16
## 390	8.44	107	8	328	35	42	12
## 391	5.47	111	9	61	67	75	12
## 392	6.10	124	0	49	56	63	16
## 393	4.53	130	13	315	34	42	13
## 394	5.57	120	10	26	30	51	17
## 395	5.35	139	19	366	33	58	16
## 396	12.57	128	17	203	33	108	14
## 397	6.14	120	3	37	55	23	11
## 398	7.41	159	12	368	40	26	18
## 399	5.94	95	7	284	50	79	12
## 400	9.71	120	0	27	49	37	16

#Plot correlation headmap

```
library(GGally)
```

```
## Warning: package 'GGally' was built under R version 4.1.2
```

```
## Registered S3 method overwritten by 'GGally':
```

```
##   method from
```

```
##   +.gg    ggplot2
```

```
ggcorr(Carseats_Filtered, label = TRUE, palette = "RdBu", name = "Correlation", hjust = 0.75, label_size = 12)
```



#Check to explore Missing Data

```
#We Look at the summary of the dataset and see if there are NA's present in variables/columns
NA_perct <- function(df, fmt = F) {
  return (df %>%
    is.na() %>%
    colMeans() %>%
    sapply(function(x) {
      if (fmt) {
        return(sprintf("%.5f%", x))
      }
      return (x)
    })
  )
}

NA_perct_df <- NA_perct(Carseats_Filtered) %>%
  data_frame(Colomns = names(.), `NA %` = .) %>%
  mutate_at(
    vars(`NA %`),
    funs(round(. * 100, 2))
  ) %>%
  mutate(label = sprintf("%g%%", `NA %`)) %>%
  arrange(desc(`NA %`))
```

```
## Warning: 'data_frame()' was deprecated in tibble 1.1.0.
## Please use 'tibble()' instead.
```

```
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_warnings()' to see where this warning was generated.

## Warning: 'funs()' was deprecated in dplyr 0.8.0.
## Please use a list of either functions or lambdas:
##
##   # Simple named list:
##   list(mean = mean, median = median)
##
##   # Auto named with 'tibble::lst()':
##   tibble::lst(mean, median)
##
##   # Using lambdas
##   list(~ mean(., trim = .2), ~ median(., na.rm = TRUE))
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_warnings()' to see where this warning was generated.
```

```
NA_perct_df %>% select(-label)
```

```
## # A tibble: 7 x 2
##   Columns      'NA %'
##   <chr>        <dbl>
## 1 i..Sales      0
## 2 Price         0
## 3 Advertising   0
## 4 Population    0
## 5 Age           0
## 6 Income        0
## 7 Education     0
```

This indicates that there are no missing data

```
summary(Carseats_Filtered)
```

```
##      i..Sales      Price      Advertising      Population
##  Min.   : 0.000   Min.    : 24.0   Min.     : 0.000   Min.     : 10.0
## 1st Qu.: 5.390   1st Qu.:100.0   1st Qu.: 0.000   1st Qu.:139.0
## Median : 7.490   Median :117.0   Median : 5.000   Median :272.0
## Mean   : 7.496   Mean    :115.8   Mean    : 6.635   Mean    :264.8
## 3rd Qu.: 9.320   3rd Qu.:131.0   3rd Qu.:12.000   3rd Qu.:398.5
## Max.   :16.270   Max.     :191.0   Max.     :29.000   Max.     :509.0
##      Age      Income      Education
##  Min.   :25.00   Min.    : 21.00   Min.     :10.0
## 1st Qu.:39.75   1st Qu.: 42.75   1st Qu.:12.0
## Median :54.50   Median : 69.00   Median :14.0
## Mean   :53.32   Mean    : 68.66   Mean     :13.9
## 3rd Qu.:66.00   3rd Qu.: 91.00   3rd Qu.:16.0
## Max.   :80.00   Max.     :120.00   Max.     :18.0
```

```
glimpse(Carseats_Filtered)
```

```
## Rows: 400
## Columns: 7
## $ i..Sales    <dbl> 9.50, 11.22, 10.06, 7.40, 4.15, 10.81, 6.63, 11.85, 6.54, ~
## $ Price       <int> 120, 83, 80, 97, 128, 72, 108, 120, 124, 124, 100, 94, 136~
## $ Advertising <int> 11, 16, 10, 4, 3, 13, 0, 15, 0, 0, 9, 4, 2, 11, 11, 5, 0, ~
## $ Population  <int> 276, 260, 269, 466, 340, 501, 45, 425, 108, 131, 150, 503,~
## $ Age         <int> 42, 65, 59, 55, 38, 78, 71, 67, 76, 76, 26, 50, 62, 53, 52~
## $ Income      <int> 73, 48, 35, 100, 64, 113, 105, 81, 110, 113, 78, 94, 35, 2~
## $ Education   <int> 17, 10, 12, 14, 13, 16, 15, 10, 10, 17, 10, 13, 18, 18, 18~
```

Predicting Sales of Baby Car Seats

```
MyDecisionTree <-Carseats_Filtered[, 1:7]
Model_2 <-rpart(i..Sales~., data=Carseats_Filtered, method="anova")
summary(Model_2)
```

```
## Call:
## rpart(formula = i..Sales ~ ., data = Carseats_Filtered, method = "anova")
##      n= 400
##
##              CP nsplit rel error    xerror      xstd
## 1  0.14251535      0 1.0000000 1.0025580 0.06924325
## 2  0.08034146      1 0.8574847 0.9049745 0.06512823
## 3  0.06251702      2 0.7771432 0.8715420 0.06434565
## 4  0.02925241      3 0.7146262 0.7820490 0.05668452
## 5  0.02537341      4 0.6853738 0.7731898 0.05343946
## 6  0.02127094      5 0.6600003 0.7645757 0.05231488
## 7  0.02059174      6 0.6387294 0.7539634 0.05251800
## 8  0.01632010      7 0.6181377 0.7362951 0.05084472
## 9  0.01521801      8 0.6018176 0.7583298 0.05538272
## 10 0.01042023      9 0.5865996 0.8027608 0.05777675
## 11 0.01000559     10 0.5761793 0.8120555 0.05809388
## 12 0.01000000     12 0.5561681 0.8171755 0.05885499
##
## Variable importance
##      Price Advertising      Age      Income Population  Education
##      49           18        16           8           6           3
##
## Node number 1: 400 observations,      complexity param=0.1425153
##      mean=7.496325, MSE=7.955687
##      left son=2 (329 obs) right son=3 (71 obs)
##      Primary splits:
##      Price      < 94.5  to the right, improve=0.14251530, (0 missing)
##      Advertising < 7.5   to the left, improve=0.07303226, (0 missing)
##      Age         < 61.5  to the right, improve=0.07120203, (0 missing)
##      Income      < 61.5  to the left, improve=0.02840494, (0 missing)
##      Population  < 174.5 to the left, improve=0.01077467, (0 missing)
##
## Node number 2: 329 observations,      complexity param=0.08034146
```

```

## mean=7.001672, MSE=6.815199
## left son=4 (174 obs) right son=5 (155 obs)
## Primary splits:
## Advertising < 6.5 to the left, improve=0.11402580, (0 missing)
## Price < 136.5 to the right, improve=0.08411056, (0 missing)
## Age < 63.5 to the right, improve=0.08091745, (0 missing)
## Income < 60.5 to the left, improve=0.03394126, (0 missing)
## Population < 23 to the left, improve=0.01831455, (0 missing)
## Surrogate splits:
## Population < 223 to the left, agree=0.599, adj=0.148, (0 split)
## Education < 10.5 to the right, agree=0.565, adj=0.077, (0 split)
## Age < 53.5 to the right, agree=0.547, adj=0.039, (0 split)
## Income < 114.5 to the left, agree=0.547, adj=0.039, (0 split)
## Price < 106.5 to the right, agree=0.544, adj=0.032, (0 split)
##
## Node number 3: 71 observations, complexity param=0.02537341
## mean=9.788451, MSE=6.852836
## left son=6 (36 obs) right son=7 (35 obs)
## Primary splits:
## Age < 54.5 to the right, improve=0.16595410, (0 missing)
## Price < 75.5 to the right, improve=0.08365773, (0 missing)
## Income < 30.5 to the left, improve=0.03322169, (0 missing)
## Education < 10.5 to the right, improve=0.03019634, (0 missing)
## Population < 268.5 to the left, improve=0.02383306, (0 missing)
## Surrogate splits:
## Advertising < 4.5 to the right, agree=0.606, adj=0.200, (0 split)
## Price < 73 to the right, agree=0.592, adj=0.171, (0 split)
## Population < 272.5 to the left, agree=0.592, adj=0.171, (0 split)
## Income < 79.5 to the right, agree=0.592, adj=0.171, (0 split)
## Education < 11.5 to the left, agree=0.577, adj=0.143, (0 split)
##
## Node number 4: 174 observations, complexity param=0.02127094
## mean=6.169655, MSE=4.942347
## left son=8 (58 obs) right son=9 (116 obs)
## Primary splits:
## Age < 63.5 to the right, improve=0.078712160, (0 missing)
## Price < 130.5 to the right, improve=0.048919280, (0 missing)
## Population < 26.5 to the left, improve=0.030421540, (0 missing)
## Income < 67.5 to the left, improve=0.027749670, (0 missing)
## Advertising < 0.5 to the left, improve=0.006795377, (0 missing)
## Surrogate splits:
## Income < 22.5 to the left, agree=0.678, adj=0.034, (0 split)
## Price < 96.5 to the left, agree=0.672, adj=0.017, (0 split)
## Population < 26.5 to the left, agree=0.672, adj=0.017, (0 split)
##
## Node number 5: 155 observations, complexity param=0.06251702
## mean=7.935677, MSE=7.268151
## left son=10 (28 obs) right son=11 (127 obs)
## Primary splits:
## Price < 136.5 to the right, improve=0.17659580, (0 missing)
## Age < 73.5 to the right, improve=0.08000201, (0 missing)
## Income < 60.5 to the left, improve=0.05360755, (0 missing)
## Advertising < 13.5 to the left, improve=0.03920507, (0 missing)
## Population < 399 to the left, improve=0.01037956, (0 missing)

```

```

## Surrogate splits:
## Advertising < 24.5 to the right, agree=0.826, adj=0.036, (0 split)
##
## Node number 6: 36 observations, complexity param=0.0163201
## mean=8.736944, MSE=4.961043
## left son=12 (12 obs) right son=13 (24 obs)
## Primary splits:
## Price < 89.5 to the right, improve=0.29079360, (0 missing)
## Income < 39.5 to the left, improve=0.19043350, (0 missing)
## Advertising < 11.5 to the left, improve=0.17891930, (0 missing)
## Age < 75.5 to the right, improve=0.04316067, (0 missing)
## Education < 14.5 to the left, improve=0.03411396, (0 missing)
## Surrogate splits:
## Advertising < 16.5 to the right, agree=0.722, adj=0.167, (0 split)
## Income < 37.5 to the left, agree=0.722, adj=0.167, (0 split)
## Age < 56.5 to the left, agree=0.694, adj=0.083, (0 split)
##
## Node number 7: 35 observations
## mean=10.87, MSE=6.491674
##
## Node number 8: 58 observations, complexity param=0.01042023
## mean=5.287586, MSE=3.93708
## left son=16 (10 obs) right son=17 (48 obs)
## Primary splits:
## Price < 137 to the right, improve=0.14521540, (0 missing)
## Education < 15.5 to the right, improve=0.07995394, (0 missing)
## Income < 35.5 to the left, improve=0.04206708, (0 missing)
## Age < 79.5 to the left, improve=0.02799057, (0 missing)
## Population < 52.5 to the left, improve=0.01914342, (0 missing)
##
## Node number 9: 116 observations, complexity param=0.01000559
## mean=6.61069, MSE=4.861446
## left son=18 (58 obs) right son=19 (58 obs)
## Primary splits:
## Income < 67 to the left, improve=0.05085914, (0 missing)
## Population < 392 to the right, improve=0.04476721, (0 missing)
## Price < 127 to the right, improve=0.04210762, (0 missing)
## Age < 37.5 to the right, improve=0.02858424, (0 missing)
## Education < 14.5 to the left, improve=0.01187387, (0 missing)
## Surrogate splits:
## Education < 12.5 to the right, agree=0.586, adj=0.172, (0 split)
## Age < 58.5 to the left, agree=0.578, adj=0.155, (0 split)
## Price < 144.5 to the left, agree=0.569, adj=0.138, (0 split)
## Population < 479 to the right, agree=0.560, adj=0.121, (0 split)
## Advertising < 2.5 to the right, agree=0.543, adj=0.086, (0 split)
##
## Node number 10: 28 observations
## mean=5.522857, MSE=5.084213
##
## Node number 11: 127 observations, complexity param=0.02925241
## mean=8.467638, MSE=6.183142
## left son=22 (29 obs) right son=23 (98 obs)
## Primary splits:
## Age < 65.5 to the right, improve=0.11854590, (0 missing)

```

```

##      Income      < 51.5  to the left,  improve=0.08076060, (0 missing)
##      Advertising < 13.5  to the left,  improve=0.04801701, (0 missing)
##      Education   < 11.5  to the right, improve=0.02471512, (0 missing)
##      Population  < 479   to the left,  improve=0.01908657, (0 missing)
##
## Node number 12: 12 observations
##   mean=7.038333, MSE=2.886964
##
## Node number 13: 24 observations
##   mean=9.58625, MSE=3.834123
##
## Node number 16: 10 observations
##   mean=3.631, MSE=5.690169
##
## Node number 17: 48 observations
##   mean=5.632708, MSE=2.88102
##
## Node number 18: 58 observations
##   mean=6.113448, MSE=3.739109
##
## Node number 19: 58 observations,    complexity param=0.01000559
##   mean=7.107931, MSE=5.489285
##   left son=38 (10 obs) right son=39 (48 obs)
##   Primary splits:
##     Population < 390.5 to the right, improve=0.10993270, (0 missing)
##     Price      < 124.5 to the right, improve=0.07534567, (0 missing)
##     Advertising < 0.5   to the left,  improve=0.07060488, (0 missing)
##     Age        < 45.5   to the right, improve=0.04611510, (0 missing)
##     Education  < 11.5   to the right, improve=0.03722944, (0 missing)
##
## Node number 22: 29 observations
##   mean=6.893793, MSE=6.08343
##
## Node number 23: 98 observations,    complexity param=0.02059174
##   mean=8.933367, MSE=5.262759
##   left son=46 (34 obs) right son=47 (64 obs)
##   Primary splits:
##     Income      < 60.5  to the left,  improve=0.12705480, (0 missing)
##     Advertising < 13.5  to the left,  improve=0.07114001, (0 missing)
##     Price       < 118.5 to the right, improve=0.06932216, (0 missing)
##     Education   < 11.5  to the right, improve=0.03377416, (0 missing)
##     Age         < 49.5  to the right, improve=0.02289004, (0 missing)
##   Surrogate splits:
##     Education < 17.5  to the right, agree=0.663, adj=0.029, (0 split)
##
## Node number 38: 10 observations
##   mean=5.406, MSE=2.508524
##
## Node number 39: 48 observations
##   mean=7.4625, MSE=5.381106
##
## Node number 46: 34 observations,    complexity param=0.01521801
##   mean=7.811471, MSE=4.756548
##   left son=92 (19 obs) right son=93 (15 obs)

```



```

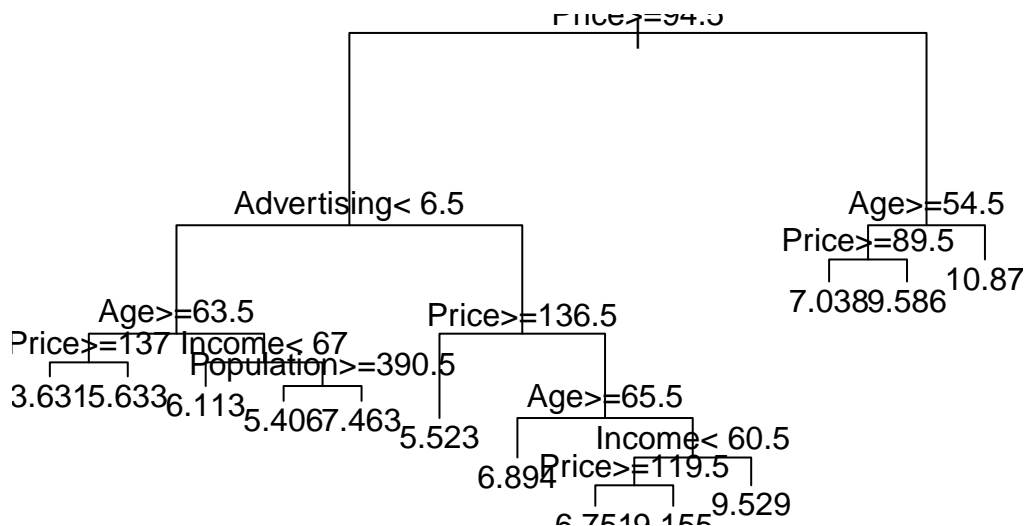
## Primary splits:
##   Price      < 119.5 to the right, improve=0.29945020, (0 missing)
##   Advertising < 11.5  to the left,  improve=0.14268440, (0 missing)
##   Income     < 40.5  to the right, improve=0.12781140, (0 missing)
##   Population < 152   to the left,  improve=0.03601768, (0 missing)
##   Age        < 49.5  to the right, improve=0.02748814, (0 missing)
## Surrogate splits:
##   Education  < 12.5  to the right, agree=0.676, adj=0.267, (0 split)
##   Advertising < 7.5   to the right, agree=0.647, adj=0.200, (0 split)
##   Age        < 53.5  to the left,  agree=0.647, adj=0.200, (0 split)
##   Population < 240   to the right, agree=0.618, adj=0.133, (0 split)
##   Income     < 41.5  to the right, agree=0.618, adj=0.133, (0 split)
##
## Node number 47: 64 observations
##   mean=9.529375, MSE=4.5078
##
## Node number 92: 19 observations
##   mean=6.751053, MSE=3.378915
##
## Node number 93: 15 observations
##   mean=9.154667, MSE=3.273025

```

```

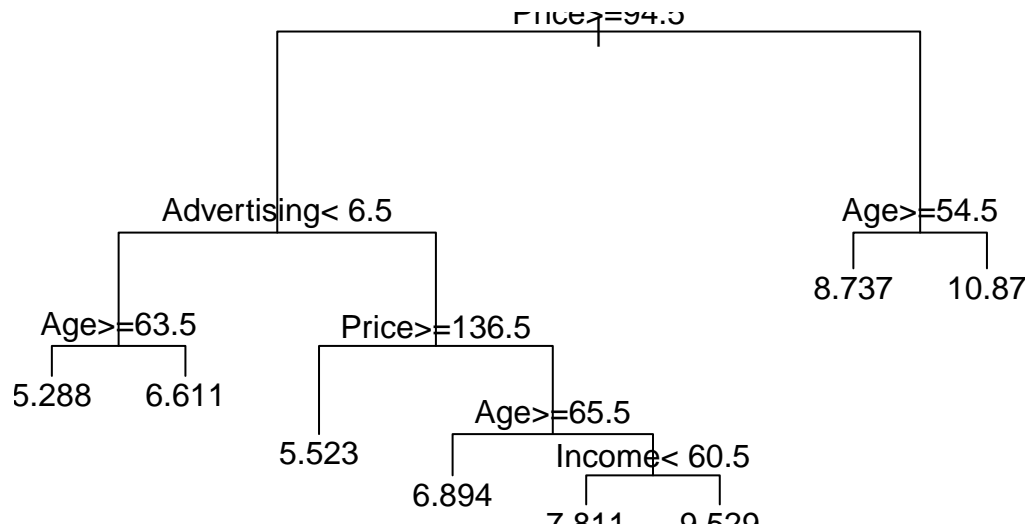
# Too Complex
plot(Model_2)
text(Model_2)

```



##Less Complex

```
Model_2 = rpart (i..Sales~.,data=Carseats_Filtered, method="anova", control = rpart.control(minsplit = 1))
plot(Model_2)
text(Model_2)
```



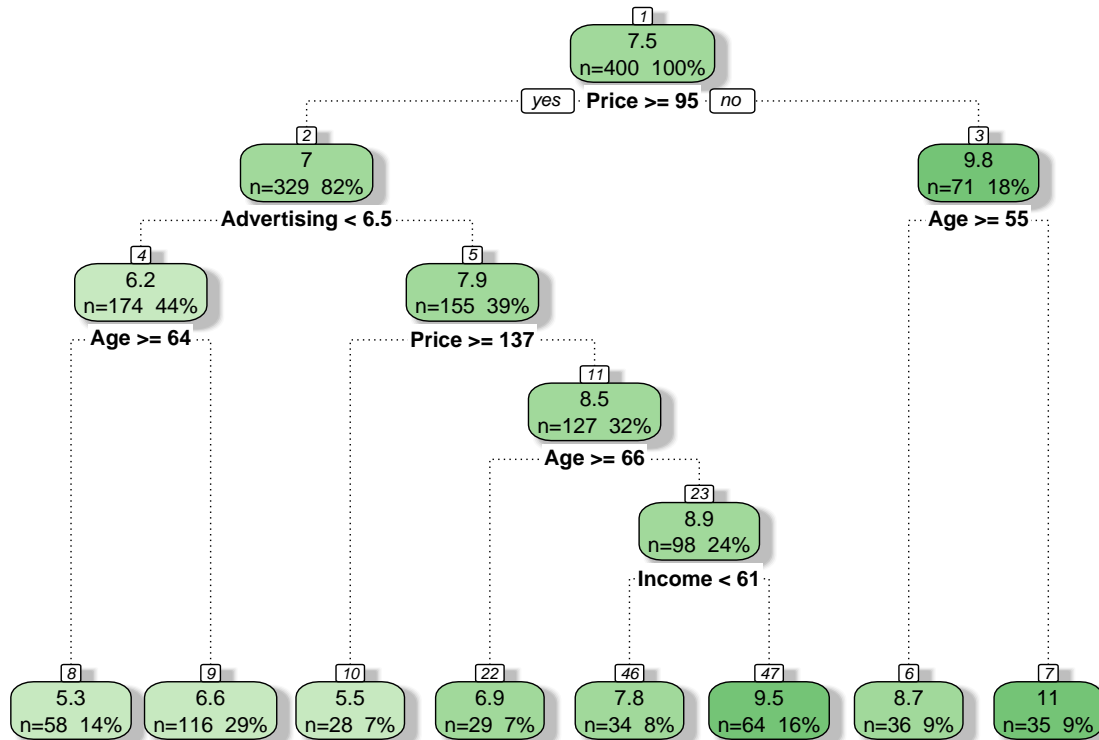
#Difficult to follow

```
print(Model_2)
```

```
## n= 400
##
## node), split, n, deviance, yval
##      * denotes terminal node
##
## 1) root 400 3182.2750  7.496325
##    2) Price>=94.5 329 2242.2000  7.001672
##      4) Advertising< 6.5 174  859.9684  6.169655
##        8) Age>=63.5 58  228.3507  5.287586 *
##        9) Age< 63.5 116  563.9277  6.610690 *
##      5) Advertising>=6.5 155 1126.5630  7.935677
##        10) Price>=136.5 28  142.3580  5.522857 *
##        11) Price< 136.5 127  785.2591  8.467638
##          22) Age>=65.5 29  176.4195  6.893793 *
##          23) Age< 65.5 98  515.7504  8.933367
##            46) Income< 60.5 34  161.7226  7.811471 *
##            47) Income>=60.5 64  288.4992  9.529375 *
```

```
##      3) Price< 94.5 71  486.5513  9.788451
##      6) Age>=54.5 36  178.5976  8.736944 *
##      7) Age< 54.5 35  227.2086 10.870000 *
```

```
library(rattle)
library(rpart.plot)
fancyRpartPlot(Model_2)
```



Rattle 2022-Apr-12 13:39:32 Mukht

##QB2. Consider the following input: Sales=9, Price=6.54, Population=124, Advertising=0, Age=76, Income= 110, Education=10 What will be the estimated Sales for this record using the decision tree model?

```
Model_3 <- data.frame(Price=6.54, Population=124, Advertising=0, Age=76, Income= 110, Education=10)
predict(Model_2, Model_3 , method = "anova")
```

```
##      1
## 8.736944
```