

## personal assignment 2

```
#(a)  
library(ISLR)  
attach(Carseats)  
?Carseats  
na.omit(Carseats)
```

##	Sales	CompPrice	Income	Advertising	Population	Price	ShelveLoc	Age
## 1	9.50	138	73	11	276	120	Bad	42
## 2	11.22	111	48	16	260	83	Good	65
## 3	10.06	113	35	10	269	80	Medium	59
## 4	7.40	117	100	4	466	97	Medium	55
## 5	4.15	141	64	3	340	128	Bad	38
## 6	10.81	124	113	13	501	72	Bad	78
## 7	6.63	115	105	0	45	108	Medium	71
## 8	11.85	136	81	15	425	120	Good	67
## 9	6.54	132	110	0	108	124	Medium	76
## 10	4.69	132	113	0	131	124	Medium	76
## 11	9.01	121	78	9	150	100	Bad	26
## 12	11.96	117	94	4	503	94	Good	50
## 13	3.98	122	35	2	393	136	Medium	62
## 14	10.96	115	28	11	29	86	Good	53
## 15	11.17	107	117	11	148	118	Good	52
## 16	8.71	149	95	5	400	144	Medium	76
## 17	7.58	118	32	0	284	110	Good	63
## 18	12.29	147	74	13	251	131	Good	52
## 19	13.91	110	110	0	408	68	Good	46
## 20	8.73	129	76	16	58	121	Medium	69
## 21	6.41	125	90	2	367	131	Medium	35
## 22	12.13	134	29	12	239	109	Good	62
## 23	5.08	128	46	6	497	138	Medium	42
## 24	5.87	121	31	0	292	109	Medium	79
## 25	10.14	145	119	16	294	113	Bad	42
## 26	14.90	139	32	0	176	82	Good	54
## 27	8.33	107	115	11	496	131	Good	50
## 28	5.27	98	118	0	19	107	Medium	64
## 29	2.99	103	74	0	359	97	Bad	55
## 30	7.81	104	99	15	226	102	Bad	58
## 31	13.55	125	94	0	447	89	Good	30
## 32	8.25	136	58	16	241	131	Medium	44
## 33	6.20	107	32	12	236	137	Good	64
## 34	8.77	114	38	13	317	128	Good	50
## 35	2.67	115	54	0	406	128	Medium	42
## 36	11.07	131	84	11	29	96	Medium	44
## 37	8.89	122	76	0	270	100	Good	60
## 38	4.95	121	41	5	412	110	Medium	54
## 39	6.59	109	73	0	454	102	Medium	65
## 40	3.24	130	60	0	144	138	Bad	38
## 41	2.07	119	98	0	18	126	Bad	73
## 42	7.96	157	53	0	403	124	Bad	58
## 43	10.43	77	69	0	25	24	Medium	50
## 44	4.12	123	42	11	16	134	Medium	59

## 45	4.16	85	79	6	325	95	Medium	69
## 46	4.56	141	63	0	168	135	Bad	44
## 47	12.44	127	90	14	16	70	Medium	48
## 48	4.38	126	98	0	173	108	Bad	55
## 49	3.91	116	52	0	349	98	Bad	69
## 50	10.61	157	93	0	51	149	Good	32
## 51	1.42	99	32	18	341	108	Bad	80
## 52	4.42	121	90	0	150	108	Bad	75
## 53	7.91	153	40	3	112	129	Bad	39
## 54	6.92	109	64	13	39	119	Medium	61
## 55	4.90	134	103	13	25	144	Medium	76
## 56	6.85	143	81	5	60	154	Medium	61
## 57	11.91	133	82	0	54	84	Medium	50
## 58	0.91	93	91	0	22	117	Bad	75
## 59	5.42	103	93	15	188	103	Bad	74
## 60	5.21	118	71	4	148	114	Medium	80
## 61	8.32	122	102	19	469	123	Bad	29
## 62	7.32	105	32	0	358	107	Medium	26
## 63	1.82	139	45	0	146	133	Bad	77
## 64	8.47	119	88	10	170	101	Medium	61
## 65	7.80	100	67	12	184	104	Medium	32
## 66	4.90	122	26	0	197	128	Medium	55
## 67	8.85	127	92	0	508	91	Medium	56
## 68	9.01	126	61	14	152	115	Medium	47
## 69	13.39	149	69	20	366	134	Good	60
## 70	7.99	127	59	0	339	99	Medium	65
## 71	9.46	89	81	15	237	99	Good	74
## 72	6.50	148	51	16	148	150	Medium	58
## 73	5.52	115	45	0	432	116	Medium	25
## 74	12.61	118	90	10	54	104	Good	31
## 75	6.20	150	68	5	125	136	Medium	64
## 76	8.55	88	111	23	480	92	Bad	36
## 77	10.64	102	87	10	346	70	Medium	64
## 78	7.70	118	71	12	44	89	Medium	67
## 79	4.43	134	48	1	139	145	Medium	65
## 80	9.14	134	67	0	286	90	Bad	41
## 81	8.01	113	100	16	353	79	Bad	68
## 82	7.52	116	72	0	237	128	Good	70
## 83	11.62	151	83	4	325	139	Good	28
## 84	4.42	109	36	7	468	94	Bad	56
## 85	2.23	111	25	0	52	121	Bad	43
## 86	8.47	125	103	0	304	112	Medium	49
## 87	8.70	150	84	9	432	134	Medium	64
## 88	11.70	131	67	7	272	126	Good	54
## 89	6.56	117	42	7	144	111	Medium	62
## 90	7.95	128	66	3	493	119	Medium	45
## 91	5.33	115	22	0	491	103	Medium	64
## 92	4.81	97	46	11	267	107	Medium	80
## 93	4.53	114	113	0	97	125	Medium	29
## 94	8.86	145	30	0	67	104	Medium	55
## 95	8.39	115	97	5	134	84	Bad	55
## 96	5.58	134	25	10	237	148	Medium	59
## 97	9.48	147	42	10	407	132	Good	73
## 98	7.45	161	82	5	287	129	Bad	33

## 99	12.49	122	77	24	382	127	Good	36
## 100	4.88	121	47	3	220	107	Bad	56
## 101	4.11	113	69	11	94	106	Medium	76
## 102	6.20	128	93	0	89	118	Medium	34
## 103	5.30	113	22	0	57	97	Medium	65
## 104	5.07	123	91	0	334	96	Bad	78
## 105	4.62	121	96	0	472	138	Medium	51
## 106	5.55	104	100	8	398	97	Medium	61
## 107	0.16	102	33	0	217	139	Medium	70
## 108	8.55	134	107	0	104	108	Medium	60
## 109	3.47	107	79	2	488	103	Bad	65
## 110	8.98	115	65	0	217	90	Medium	60
## 111	9.00	128	62	7	125	116	Medium	43
## 112	6.62	132	118	12	272	151	Medium	43
## 113	6.67	116	99	5	298	125	Good	62
## 114	6.01	131	29	11	335	127	Bad	33
## 115	9.31	122	87	9	17	106	Medium	65
## 116	8.54	139	35	0	95	129	Medium	42
## 117	5.08	135	75	0	202	128	Medium	80
## 118	8.80	145	53	0	507	119	Medium	41
## 119	7.57	112	88	2	243	99	Medium	62
## 120	7.37	130	94	8	137	128	Medium	64
## 121	6.87	128	105	11	249	131	Medium	63
## 122	11.67	125	89	10	380	87	Bad	28
## 123	6.88	119	100	5	45	108	Medium	75
## 124	8.19	127	103	0	125	155	Good	29
## 125	8.87	131	113	0	181	120	Good	63
## 126	9.34	89	78	0	181	49	Medium	43
## 127	11.27	153	68	2	60	133	Good	59
## 128	6.52	125	48	3	192	116	Medium	51
## 129	4.96	133	100	3	350	126	Bad	55
## 130	4.47	143	120	7	279	147	Bad	40
## 131	8.41	94	84	13	497	77	Medium	51
## 132	6.50	108	69	3	208	94	Medium	77
## 133	9.54	125	87	9	232	136	Good	72
## 134	7.62	132	98	2	265	97	Bad	62
## 135	3.67	132	31	0	327	131	Medium	76
## 136	6.44	96	94	14	384	120	Medium	36
## 137	5.17	131	75	0	10	120	Bad	31
## 138	6.52	128	42	0	436	118	Medium	80
## 139	10.27	125	103	12	371	109	Medium	44
## 140	12.30	146	62	10	310	94	Medium	30
## 141	6.03	133	60	10	277	129	Medium	45
## 142	6.53	140	42	0	331	131	Bad	28
## 143	7.44	124	84	0	300	104	Medium	77
## 144	0.53	122	88	7	36	159	Bad	28
## 145	9.09	132	68	0	264	123	Good	34
## 146	8.77	144	63	11	27	117	Medium	47
## 147	3.90	114	83	0	412	131	Bad	39
## 148	10.51	140	54	9	402	119	Good	41
## 149	7.56	110	119	0	384	97	Medium	72
## 150	11.48	121	120	13	140	87	Medium	56
## 151	10.49	122	84	8	176	114	Good	57
## 152	10.77	111	58	17	407	103	Good	75

## 153	7.64	128	78	0	341	128	Good	45
## 154	5.93	150	36	7	488	150	Medium	25
## 155	6.89	129	69	10	289	110	Medium	50
## 156	7.71	98	72	0	59	69	Medium	65
## 157	7.49	146	34	0	220	157	Good	51
## 158	10.21	121	58	8	249	90	Medium	48
## 159	12.53	142	90	1	189	112	Good	39
## 160	9.32	119	60	0	372	70	Bad	30
## 161	4.67	111	28	0	486	111	Medium	29
## 162	2.93	143	21	5	81	160	Medium	67
## 163	3.63	122	74	0	424	149	Medium	51
## 164	5.68	130	64	0	40	106	Bad	39
## 165	8.22	148	64	0	58	141	Medium	27
## 166	0.37	147	58	7	100	191	Bad	27
## 167	6.71	119	67	17	151	137	Medium	55
## 168	6.71	106	73	0	216	93	Medium	60
## 169	7.30	129	89	0	425	117	Medium	45
## 170	11.48	104	41	15	492	77	Good	73
## 171	8.01	128	39	12	356	118	Medium	71
## 172	12.49	93	106	12	416	55	Medium	75
## 173	9.03	104	102	13	123	110	Good	35
## 174	6.38	135	91	5	207	128	Medium	66
## 175	0.00	139	24	0	358	185	Medium	79
## 176	7.54	115	89	0	38	122	Medium	25
## 177	5.61	138	107	9	480	154	Medium	47
## 178	10.48	138	72	0	148	94	Medium	27
## 179	10.66	104	71	14	89	81	Medium	25
## 180	7.78	144	25	3	70	116	Medium	77
## 181	4.94	137	112	15	434	149	Bad	66
## 182	7.43	121	83	0	79	91	Medium	68
## 183	4.74	137	60	4	230	140	Bad	25
## 184	5.32	118	74	6	426	102	Medium	80
## 185	9.95	132	33	7	35	97	Medium	60
## 186	10.07	130	100	11	449	107	Medium	64
## 187	8.68	120	51	0	93	86	Medium	46
## 188	6.03	117	32	0	142	96	Bad	62
## 189	8.07	116	37	0	426	90	Medium	76
## 190	12.11	118	117	18	509	104	Medium	26
## 191	8.79	130	37	13	297	101	Medium	37
## 192	6.67	156	42	13	170	173	Good	74
## 193	7.56	108	26	0	408	93	Medium	56
## 194	13.28	139	70	7	71	96	Good	61
## 195	7.23	112	98	18	481	128	Medium	45
## 196	4.19	117	93	4	420	112	Bad	66
## 197	4.10	130	28	6	410	133	Bad	72
## 198	2.52	124	61	0	333	138	Medium	76
## 199	3.62	112	80	5	500	128	Medium	69
## 200	6.42	122	88	5	335	126	Medium	64
## 201	5.56	144	92	0	349	146	Medium	62
## 202	5.94	138	83	0	139	134	Medium	54
## 203	4.10	121	78	4	413	130	Bad	46
## 204	2.05	131	82	0	132	157	Bad	25
## 205	8.74	155	80	0	237	124	Medium	37
## 206	5.68	113	22	1	317	132	Medium	28

## 207	4.97	162	67	0	27	160	Medium	77
## 208	8.19	111	105	0	466	97	Bad	61
## 209	7.78	86	54	0	497	64	Bad	33
## 210	3.02	98	21	11	326	90	Bad	76
## 211	4.36	125	41	2	357	123	Bad	47
## 212	9.39	117	118	14	445	120	Medium	32
## 213	12.04	145	69	19	501	105	Medium	45
## 214	8.23	149	84	5	220	139	Medium	33
## 215	4.83	115	115	3	48	107	Medium	73
## 216	2.34	116	83	15	170	144	Bad	71
## 217	5.73	141	33	0	243	144	Medium	34
## 218	4.34	106	44	0	481	111	Medium	70
## 219	9.70	138	61	12	156	120	Medium	25
## 220	10.62	116	79	19	359	116	Good	58
## 221	10.59	131	120	15	262	124	Medium	30
## 222	6.43	124	44	0	125	107	Medium	80
## 223	7.49	136	119	6	178	145	Medium	35
## 224	3.45	110	45	9	276	125	Medium	62
## 225	4.10	134	82	0	464	141	Medium	48
## 226	6.68	107	25	0	412	82	Bad	36
## 227	7.80	119	33	0	245	122	Good	56
## 228	8.69	113	64	10	68	101	Medium	57
## 229	5.40	149	73	13	381	163	Bad	26
## 230	11.19	98	104	0	404	72	Medium	27
## 231	5.16	115	60	0	119	114	Bad	38
## 232	8.09	132	69	0	123	122	Medium	27
## 233	13.14	137	80	10	24	105	Good	61
## 234	8.65	123	76	18	218	120	Medium	29
## 235	9.43	115	62	11	289	129	Good	56
## 236	5.53	126	32	8	95	132	Medium	50
## 237	9.32	141	34	16	361	108	Medium	69
## 238	9.62	151	28	8	499	135	Medium	48
## 239	7.36	121	24	0	200	133	Good	73
## 240	3.89	123	105	0	149	118	Bad	62
## 241	10.31	159	80	0	362	121	Medium	26
## 242	12.01	136	63	0	160	94	Medium	38
## 243	4.68	124	46	0	199	135	Medium	52
## 244	7.82	124	25	13	87	110	Medium	57
## 245	8.78	130	30	0	391	100	Medium	26
## 246	10.00	114	43	0	199	88	Good	57
## 247	6.90	120	56	20	266	90	Bad	78
## 248	5.04	123	114	0	298	151	Bad	34
## 249	5.36	111	52	0	12	101	Medium	61
## 250	5.05	125	67	0	86	117	Bad	65
## 251	9.16	137	105	10	435	156	Good	72
## 252	3.72	139	111	5	310	132	Bad	62
## 253	8.31	133	97	0	70	117	Medium	32
## 254	5.64	124	24	5	288	122	Medium	57
## 255	9.58	108	104	23	353	129	Good	37
## 256	7.71	123	81	8	198	81	Bad	80
## 257	4.20	147	40	0	277	144	Medium	73
## 258	8.67	125	62	14	477	112	Medium	80
## 259	3.47	108	38	0	251	81	Bad	72
## 260	5.12	123	36	10	467	100	Bad	74

## 261	7.67	129	117	8	400	101	Bad	36
## 262	5.71	121	42	4	188	118	Medium	54
## 263	6.37	120	77	15	86	132	Medium	48
## 264	7.77	116	26	6	434	115	Medium	25
## 265	6.95	128	29	5	324	159	Good	31
## 266	5.31	130	35	10	402	129	Bad	39
## 267	9.10	128	93	12	343	112	Good	73
## 268	5.83	134	82	7	473	112	Bad	51
## 269	6.53	123	57	0	66	105	Medium	39
## 270	5.01	159	69	0	438	166	Medium	46
## 271	11.99	119	26	0	284	89	Good	26
## 272	4.55	111	56	0	504	110	Medium	62
## 273	12.98	113	33	0	14	63	Good	38
## 274	10.04	116	106	8	244	86	Medium	58
## 275	7.22	135	93	2	67	119	Medium	34
## 276	6.67	107	119	11	210	132	Medium	53
## 277	6.93	135	69	14	296	130	Medium	73
## 278	7.80	136	48	12	326	125	Medium	36
## 279	7.22	114	113	2	129	151	Good	40
## 280	3.42	141	57	13	376	158	Medium	64
## 281	2.86	121	86	10	496	145	Bad	51
## 282	11.19	122	69	7	303	105	Good	45
## 283	7.74	150	96	0	80	154	Good	61
## 284	5.36	135	110	0	112	117	Medium	80
## 285	6.97	106	46	11	414	96	Bad	79
## 286	7.60	146	26	11	261	131	Medium	39
## 287	7.53	117	118	11	429	113	Medium	67
## 288	6.88	95	44	4	208	72	Bad	44
## 289	6.98	116	40	0	74	97	Medium	76
## 290	8.75	143	77	25	448	156	Medium	43
## 291	9.49	107	111	14	400	103	Medium	41
## 292	6.64	118	70	0	106	89	Bad	39
## 293	11.82	113	66	16	322	74	Good	76
## 294	11.28	123	84	0	74	89	Good	59
## 295	12.66	148	76	3	126	99	Good	60
## 296	4.21	118	35	14	502	137	Medium	79
## 297	8.21	127	44	13	160	123	Good	63
## 298	3.07	118	83	13	276	104	Bad	75
## 299	10.98	148	63	0	312	130	Good	63
## 300	9.40	135	40	17	497	96	Medium	54
## 301	8.57	116	78	1	158	99	Medium	45
## 302	7.41	99	93	0	198	87	Medium	57
## 303	5.28	108	77	13	388	110	Bad	74
## 304	10.01	133	52	16	290	99	Medium	43
## 305	11.93	123	98	12	408	134	Good	29
## 306	8.03	115	29	26	394	132	Medium	33
## 307	4.78	131	32	1	85	133	Medium	48
## 308	5.90	138	92	0	13	120	Bad	61
## 309	9.24	126	80	19	436	126	Medium	52
## 310	11.18	131	111	13	33	80	Bad	68
## 311	9.53	175	65	29	419	166	Medium	53
## 312	6.15	146	68	12	328	132	Bad	51
## 313	6.80	137	117	5	337	135	Bad	38
## 314	9.33	103	81	3	491	54	Medium	66

## 315	7.72	133	33	10	333	129	Good	71
## 316	6.39	131	21	8	220	171	Good	29
## 317	15.63	122	36	5	369	72	Good	35
## 318	6.41	142	30	0	472	136	Good	80
## 319	10.08	116	72	10	456	130	Good	41
## 320	6.97	127	45	19	459	129	Medium	57
## 321	5.86	136	70	12	171	152	Medium	44
## 322	7.52	123	39	5	499	98	Medium	34
## 323	9.16	140	50	10	300	139	Good	60
## 324	10.36	107	105	18	428	103	Medium	34
## 325	2.66	136	65	4	133	150	Bad	53
## 326	11.70	144	69	11	131	104	Medium	47
## 327	4.69	133	30	0	152	122	Medium	53
## 328	6.23	112	38	17	316	104	Medium	80
## 329	3.15	117	66	1	65	111	Bad	55
## 330	11.27	100	54	9	433	89	Good	45
## 331	4.99	122	59	0	501	112	Bad	32
## 332	10.10	135	63	15	213	134	Medium	32
## 333	5.74	106	33	20	354	104	Medium	61
## 334	5.87	136	60	7	303	147	Medium	41
## 335	7.63	93	117	9	489	83	Bad	42
## 336	6.18	120	70	15	464	110	Medium	72
## 337	5.17	138	35	6	60	143	Bad	28
## 338	8.61	130	38	0	283	102	Medium	80
## 339	5.97	112	24	0	164	101	Medium	45
## 340	11.54	134	44	4	219	126	Good	44
## 341	7.50	140	29	0	105	91	Bad	43
## 342	7.38	98	120	0	268	93	Medium	72
## 343	7.81	137	102	13	422	118	Medium	71
## 344	5.99	117	42	10	371	121	Bad	26
## 345	8.43	138	80	0	108	126	Good	70
## 346	4.81	121	68	0	279	149	Good	79
## 347	8.97	132	107	0	144	125	Medium	33
## 348	6.88	96	39	0	161	112	Good	27
## 349	12.57	132	102	20	459	107	Good	49
## 350	9.32	134	27	18	467	96	Medium	49
## 351	8.64	111	101	17	266	91	Medium	63
## 352	10.44	124	115	16	458	105	Medium	62
## 353	13.44	133	103	14	288	122	Good	61
## 354	9.45	107	67	12	430	92	Medium	35
## 355	5.30	133	31	1	80	145	Medium	42
## 356	7.02	130	100	0	306	146	Good	42
## 357	3.58	142	109	0	111	164	Good	72
## 358	13.36	103	73	3	276	72	Medium	34
## 359	4.17	123	96	10	71	118	Bad	69
## 360	3.13	130	62	11	396	130	Bad	66
## 361	8.77	118	86	7	265	114	Good	52
## 362	8.68	131	25	10	183	104	Medium	56
## 363	5.25	131	55	0	26	110	Bad	79
## 364	10.26	111	75	1	377	108	Good	25
## 365	10.50	122	21	16	488	131	Good	30
## 366	6.53	154	30	0	122	162	Medium	57
## 367	5.98	124	56	11	447	134	Medium	53
## 368	14.37	95	106	0	256	53	Good	52

## 369	10.71	109	22	10	348	79	Good	74
## 370	10.26	135	100	22	463	122	Medium	36
## 371	7.68	126	41	22	403	119	Bad	42
## 372	9.08	152	81	0	191	126	Medium	54
## 373	7.80	121	50	0	508	98	Medium	65
## 374	5.58	137	71	0	402	116	Medium	78
## 375	9.44	131	47	7	90	118	Medium	47
## 376	7.90	132	46	4	206	124	Medium	73
## 377	16.27	141	60	19	319	92	Good	44
## 378	6.81	132	61	0	263	125	Medium	41
## 379	6.11	133	88	3	105	119	Medium	79
## 380	5.81	125	111	0	404	107	Bad	54
## 381	9.64	106	64	10	17	89	Medium	68
## 382	3.90	124	65	21	496	151	Bad	77
## 383	4.95	121	28	19	315	121	Medium	66
## 384	9.35	98	117	0	76	68	Medium	63
## 385	12.85	123	37	15	348	112	Good	28
## 386	5.87	131	73	13	455	132	Medium	62
## 387	5.32	152	116	0	170	160	Medium	39
## 388	8.67	142	73	14	238	115	Medium	73
## 389	8.14	135	89	11	245	78	Bad	79
## 390	8.44	128	42	8	328	107	Medium	35
## 391	5.47	108	75	9	61	111	Medium	67
## 392	6.10	153	63	0	49	124	Bad	56
## 393	4.53	129	42	13	315	130	Bad	34
## 394	5.57	109	51	10	26	120	Medium	30
## 395	5.35	130	58	19	366	139	Bad	33
## 396	12.57	138	108	17	203	128	Good	33
## 397	6.14	139	23	3	37	120	Medium	55
## 398	7.41	162	26	12	368	159	Medium	40
## 399	5.94	100	79	7	284	95	Bad	50
## 400	9.71	134	37	0	27	120	Good	49
##	Education	Urban	US					
## 1	17	Yes	Yes					
## 2	10	Yes	Yes					
## 3	12	Yes	Yes					
## 4	14	Yes	Yes					
## 5	13	Yes	No					
## 6	16	No	Yes					
## 7	15	Yes	No					
## 8	10	Yes	Yes					
## 9	10	No	No					
## 10	17	No	Yes					
## 11	10	No	Yes					
## 12	13	Yes	Yes					
## 13	18	Yes	No					
## 14	18	Yes	Yes					
## 15	18	Yes	Yes					
## 16	18	No	No					
## 17	13	Yes	No					
## 18	10	Yes	Yes					
## 19	17	No	Yes					
## 20	12	Yes	Yes					
## 21	18	Yes	Yes					



## 22	18	No	Yes
## 23	13	Yes	No
## 24	10	Yes	No
## 25	12	Yes	Yes
## 26	11	No	No
## 27	11	No	Yes
## 28	17	Yes	No
## 29	11	Yes	Yes
## 30	17	Yes	Yes
## 31	12	Yes	No
## 32	18	Yes	Yes
## 33	10	No	Yes
## 34	16	Yes	Yes
## 35	17	Yes	Yes
## 36	17	No	Yes
## 37	18	No	No
## 38	10	Yes	Yes
## 39	15	Yes	No
## 40	10	No	No
## 41	17	No	No
## 42	16	Yes	No
## 43	18	Yes	No
## 44	13	Yes	Yes
## 45	13	Yes	Yes
## 46	12	Yes	Yes
## 47	15	No	Yes
## 48	16	Yes	No
## 49	18	Yes	No
## 50	17	Yes	No
## 51	16	Yes	Yes
## 52	16	Yes	No
## 53	18	Yes	Yes
## 54	17	Yes	Yes
## 55	17	No	Yes
## 56	18	Yes	Yes
## 57	17	Yes	No
## 58	11	Yes	No
## 59	16	Yes	Yes
## 60	13	Yes	No
## 61	13	Yes	Yes
## 62	13	No	No
## 63	17	Yes	Yes
## 64	13	Yes	Yes
## 65	16	No	Yes
## 66	13	No	No
## 67	18	Yes	No
## 68	16	Yes	Yes
## 69	13	Yes	Yes
## 70	12	Yes	No
## 71	12	Yes	Yes
## 72	17	No	Yes
## 73	15	Yes	No
## 74	11	No	Yes
## 75	13	No	Yes

## 76	16	No	Yes
## 77	15	Yes	Yes
## 78	18	No	Yes
## 79	12	Yes	Yes
## 80	13	Yes	No
## 81	11	Yes	Yes
## 82	13	Yes	No
## 83	17	Yes	Yes
## 84	11	Yes	Yes
## 85	18	No	No
## 86	13	No	No
## 87	15	Yes	No
## 88	16	No	Yes
## 89	10	Yes	Yes
## 90	16	No	No
## 91	11	No	No
## 92	15	Yes	Yes
## 93	12	Yes	No
## 94	17	Yes	No
## 95	11	Yes	Yes
## 96	13	Yes	Yes
## 97	16	No	Yes
## 98	16	Yes	Yes
## 99	16	No	Yes
## 100	16	No	Yes
## 101	12	No	Yes
## 102	18	Yes	No
## 103	16	No	No
## 104	17	Yes	Yes
## 105	12	Yes	No
## 106	11	Yes	Yes
## 107	18	No	No
## 108	12	Yes	No
## 109	16	Yes	No
## 110	17	No	No
## 111	14	Yes	Yes
## 112	14	Yes	Yes
## 113	12	Yes	Yes
## 114	12	Yes	Yes
## 115	13	Yes	Yes
## 116	13	Yes	No
## 117	10	No	No
## 118	12	Yes	No
## 119	11	Yes	Yes
## 120	12	Yes	Yes
## 121	13	Yes	Yes
## 122	10	Yes	Yes
## 123	10	Yes	Yes
## 124	15	No	Yes
## 125	14	Yes	No
## 126	15	No	No
## 127	16	Yes	Yes
## 128	14	Yes	Yes
## 129	13	Yes	Yes

## 130	10	No	Yes
## 131	12	Yes	Yes
## 132	16	Yes	No
## 133	10	Yes	Yes
## 134	12	Yes	Yes
## 135	16	Yes	No
## 136	18	No	Yes
## 137	18	No	No
## 138	11	Yes	No
## 139	10	Yes	Yes
## 140	13	No	Yes
## 141	18	Yes	Yes
## 142	15	Yes	No
## 143	15	Yes	No
## 144	17	Yes	Yes
## 145	11	No	No
## 146	17	Yes	Yes
## 147	14	Yes	No
## 148	16	No	Yes
## 149	14	No	Yes
## 150	11	Yes	Yes
## 151	10	No	Yes
## 152	17	No	Yes
## 153	13	No	No
## 154	17	No	Yes
## 155	16	No	Yes
## 156	16	Yes	No
## 157	16	Yes	No
## 158	13	No	Yes
## 159	10	No	Yes
## 160	18	No	No
## 161	12	No	No
## 162	12	No	Yes
## 163	13	Yes	No
## 164	17	No	No
## 165	13	No	Yes
## 166	15	Yes	Yes
## 167	11	Yes	Yes
## 168	13	Yes	No
## 169	10	Yes	No
## 170	18	Yes	Yes
## 171	10	Yes	Yes
## 172	15	Yes	Yes
## 173	16	Yes	Yes
## 174	18	Yes	Yes
## 175	15	No	No
## 176	12	Yes	No
## 177	11	No	Yes
## 178	17	Yes	Yes
## 179	14	No	Yes
## 180	18	Yes	Yes
## 181	13	Yes	Yes
## 182	11	Yes	No
## 183	13	Yes	No

## 184	18	Yes	Yes
## 185	11	No	Yes
## 186	10	Yes	Yes
## 187	17	No	No
## 188	17	Yes	No
## 189	15	Yes	No
## 190	15	No	Yes
## 191	13	No	Yes
## 192	14	Yes	Yes
## 193	14	No	No
## 194	10	Yes	Yes
## 195	11	Yes	Yes
## 196	11	Yes	Yes
## 197	16	Yes	Yes
## 198	16	Yes	No
## 199	10	Yes	Yes
## 200	14	Yes	Yes
## 201	12	No	No
## 202	18	Yes	No
## 203	10	No	Yes
## 204	14	Yes	No
## 205	14	Yes	No
## 206	12	Yes	No
## 207	17	Yes	Yes
## 208	10	No	No
## 209	12	Yes	No
## 210	11	No	Yes
## 211	14	No	Yes
## 212	15	Yes	Yes
## 213	11	Yes	Yes
## 214	10	Yes	Yes
## 215	18	Yes	Yes
## 216	11	Yes	Yes
## 217	17	Yes	No
## 218	14	No	No
## 219	14	Yes	Yes
## 220	17	Yes	Yes
## 221	10	Yes	Yes
## 222	11	Yes	No
## 223	13	Yes	Yes
## 224	14	Yes	Yes
## 225	13	No	No
## 226	14	Yes	No
## 227	14	Yes	No
## 228	16	Yes	Yes
## 229	11	No	Yes
## 230	18	No	No
## 231	14	No	No
## 232	11	No	No
## 233	15	Yes	Yes
## 234	14	No	Yes
## 235	16	No	Yes
## 236	17	Yes	Yes
## 237	10	Yes	Yes

## 238	10	Yes	Yes
## 239	13	Yes	No
## 240	16	Yes	Yes
## 241	18	Yes	No
## 242	12	Yes	No
## 243	14	No	No
## 244	10	Yes	Yes
## 245	18	Yes	No
## 246	10	No	Yes
## 247	18	Yes	Yes
## 248	16	Yes	No
## 249	11	Yes	Yes
## 250	11	Yes	No
## 251	14	Yes	Yes
## 252	13	Yes	Yes
## 253	16	Yes	No
## 254	12	No	Yes
## 255	17	Yes	Yes
## 256	15	Yes	Yes
## 257	10	Yes	No
## 258	13	Yes	Yes
## 259	14	No	No
## 260	11	No	Yes
## 261	10	Yes	Yes
## 262	15	Yes	Yes
## 263	18	Yes	Yes
## 264	17	Yes	Yes
## 265	15	Yes	Yes
## 266	17	Yes	Yes
## 267	17	No	Yes
## 268	12	No	Yes
## 269	11	Yes	No
## 270	17	Yes	No
## 271	10	Yes	No
## 272	16	Yes	No
## 273	12	Yes	No
## 274	12	Yes	Yes
## 275	11	Yes	Yes
## 276	11	Yes	Yes
## 277	15	Yes	Yes
## 278	16	Yes	Yes
## 279	15	No	Yes
## 280	18	Yes	Yes
## 281	10	Yes	Yes
## 282	16	No	Yes
## 283	11	Yes	No
## 284	16	No	No
## 285	17	No	No
## 286	10	Yes	Yes
## 287	18	No	Yes
## 288	17	Yes	Yes
## 289	15	No	No
## 290	17	Yes	Yes
## 291	11	No	Yes

## 292	17	Yes	No
## 293	15	Yes	Yes
## 294	10	Yes	No
## 295	11	Yes	Yes
## 296	10	No	Yes
## 297	18	Yes	Yes
## 298	10	Yes	Yes
## 299	15	Yes	No
## 300	17	No	Yes
## 301	11	Yes	Yes
## 302	16	Yes	Yes
## 303	14	Yes	Yes
## 304	11	Yes	Yes
## 305	10	Yes	Yes
## 306	13	Yes	Yes
## 307	12	Yes	Yes
## 308	12	Yes	No
## 309	10	Yes	Yes
## 310	18	Yes	Yes
## 311	12	Yes	Yes
## 312	14	Yes	Yes
## 313	10	Yes	Yes
## 314	13	Yes	No
## 315	14	Yes	Yes
## 316	14	Yes	Yes
## 317	10	Yes	Yes
## 318	15	No	No
## 319	14	No	Yes
## 320	11	No	Yes
## 321	18	Yes	Yes
## 322	15	Yes	No
## 323	15	Yes	Yes
## 324	12	Yes	Yes
## 325	13	Yes	Yes
## 326	11	Yes	Yes
## 327	17	Yes	No
## 328	16	Yes	Yes
## 329	11	Yes	Yes
## 330	12	Yes	Yes
## 331	14	No	No
## 332	10	Yes	Yes
## 333	12	Yes	Yes
## 334	10	Yes	Yes
## 335	13	Yes	Yes
## 336	15	Yes	Yes
## 337	18	Yes	No
## 338	15	Yes	No
## 339	11	Yes	No
## 340	15	Yes	Yes
## 341	16	Yes	No
## 342	10	No	No
## 343	10	No	Yes
## 344	14	Yes	Yes
## 345	13	No	Yes

## 346	12	Yes	No
## 347	13	No	No
## 348	14	No	No
## 349	11	Yes	Yes
## 350	14	No	Yes
## 351	17	No	Yes
## 352	16	No	Yes
## 353	17	Yes	Yes
## 354	12	No	Yes
## 355	18	Yes	Yes
## 356	11	Yes	No
## 357	12	Yes	No
## 358	15	Yes	Yes
## 359	11	Yes	Yes
## 360	14	Yes	Yes
## 361	15	No	Yes
## 362	15	No	Yes
## 363	12	Yes	Yes
## 364	12	Yes	No
## 365	14	Yes	Yes
## 366	17	No	No
## 367	12	No	Yes
## 368	17	Yes	No
## 369	14	No	Yes
## 370	14	Yes	Yes
## 371	12	Yes	Yes
## 372	16	Yes	No
## 373	11	No	No
## 374	17	Yes	No
## 375	12	Yes	Yes
## 376	11	Yes	No
## 377	11	Yes	Yes
## 378	12	No	No
## 379	12	Yes	Yes
## 380	15	Yes	No
## 381	17	Yes	Yes
## 382	13	Yes	Yes
## 383	14	Yes	Yes
## 384	10	Yes	No
## 385	12	Yes	Yes
## 386	17	Yes	Yes
## 387	16	Yes	No
## 388	14	No	Yes
## 389	16	Yes	Yes
## 390	12	Yes	Yes
## 391	12	Yes	Yes
## 392	16	Yes	No
## 393	13	Yes	Yes
## 394	17	No	Yes
## 395	16	Yes	Yes
## 396	14	Yes	Yes
## 397	11	No	Yes
## 398	18	Yes	Yes
## 399	12	Yes	Yes

```
## 400      16    Yes Yes
```

```
lm.fit=lm(Sales~Price+Urban+US)
```

```
# As above, lm.fit is a multiple regression model of sales and three predictors.
```

```
##(b)
```

```
summary(lm.fit)
```

```
##
```

```
## Call:
```

```
## lm(formula = Sales ~ Price + Urban + US)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max  
## -6.9206 -1.6220 -0.0564  1.5786  7.0581
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t value Pr(>|t|)  
## (Intercept) 13.043469   0.651012  20.036 < 2e-16 ***  
## Price       -0.054459   0.005242 -10.389 < 2e-16 ***  
## UrbanYes    -0.021916   0.271650  -0.081  0.936  
## USYes       1.200573   0.259042   4.635 4.86e-06 ***
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
```

```
## Residual standard error: 2.472 on 396 degrees of freedom
```

```
## Multiple R-squared:  0.2393, Adjusted R-squared:  0.2335
```

```
## F-statistic: 41.52 on 3 and 396 DF,  p-value: < 2.2e-16
```

```
contrasts(US)
```

```
##      Yes
```

```
## No      0
```

```
## Yes     1
```

```
contrasts(Urban)
```

```
##      Yes
```

```
## No      0
```

```
## Yes     1
```

```
#According to the results, the coefficient for price is -0.054459, and the  
#p-value is smaller than the significance level we chose to prove it is  
#statistically significant. The number means that when the prices increase  
#an unit, the sales of the carseats for children will decrease 0.054459.  
#And the coefficients for the two qualitative predictors, which are Urban and US,  
#are -0.021916 and 1.200573 respectively. R automatically make these  
#two qualitative predictors as dummy variables, which we can know by contrasts() function.  
#Therefore, the sales will decrease if the store is in  
#the urban area. And the sale of stores in rural area is the baseline.  
#However, this correlation is not significant according to the p-value.  
#Similarly, the sales will increase if the stores are in the US,  
#and the sales outside the US are the baseline. This correlation is  
#significant from the results.
```

```
##(C)
```

```
#Sales=13.043469 - 0.054459*Price - 0.021916*Urban + 1.200537*US
```



```
##(The urban=1 if the stores are in the urban areas,otherwise it equals to 0;  
##The US=1 if the stores are in the US, otherwise it equals to 0)
```

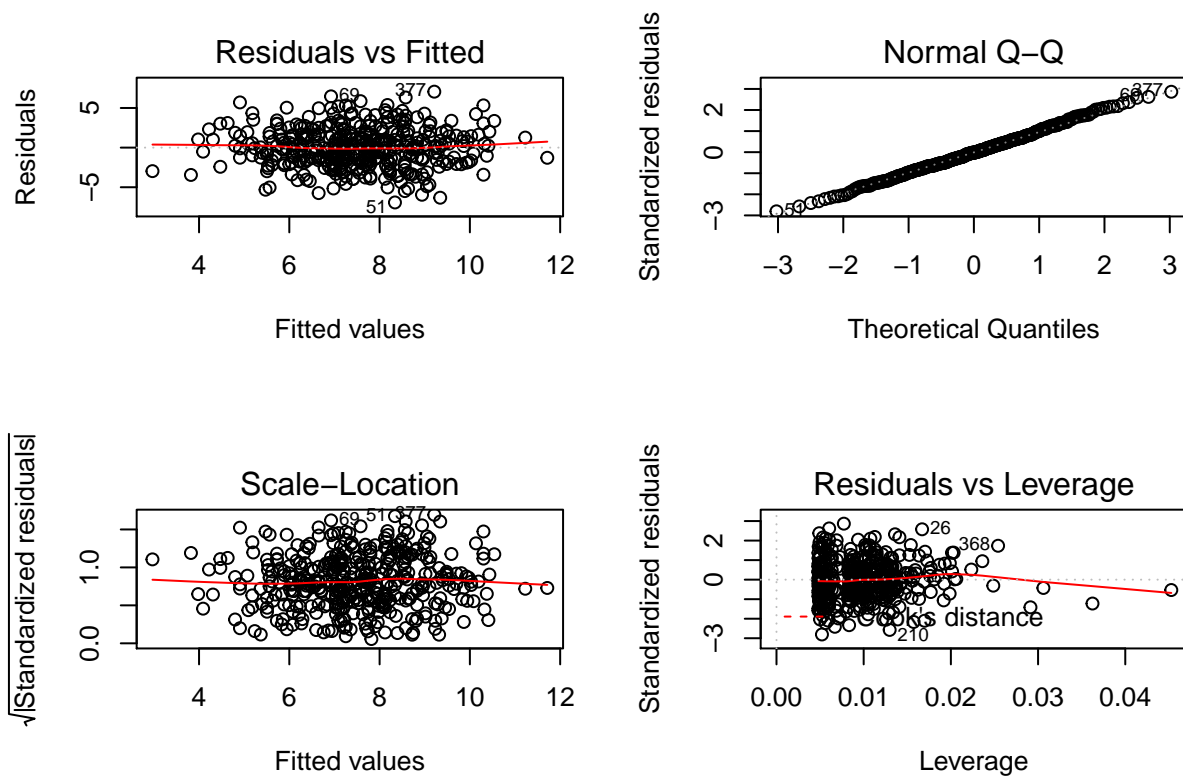
```
##(d)  
summary(lm.fit)
```

```
##  
## Call:  
## lm(formula = Sales ~ Price + Urban + US)  
##  
## Residuals:  
##      Min       1Q   Median       3Q      Max   
## -6.9206 -1.6220 -0.0564  1.5786  7.0581   
##  
## Coefficients:  
##              Estimate Std. Error t value Pr(>|t|)      
## (Intercept) 13.043469   0.651012  20.036 < 2e-16 ***  
## Price       -0.054459   0.005242 -10.389 < 2e-16 ***  
## UrbanYes    -0.021916   0.271650  -0.081  0.936      
## USYes       1.200573    0.259042   4.635 4.86e-06 ***  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 2.472 on 396 degrees of freedom  
## Multiple R-squared:  0.2393, Adjusted R-squared:  0.2335   
## F-statistic: 41.52 on 3 and 396 DF,  p-value: < 2.2e-16
```

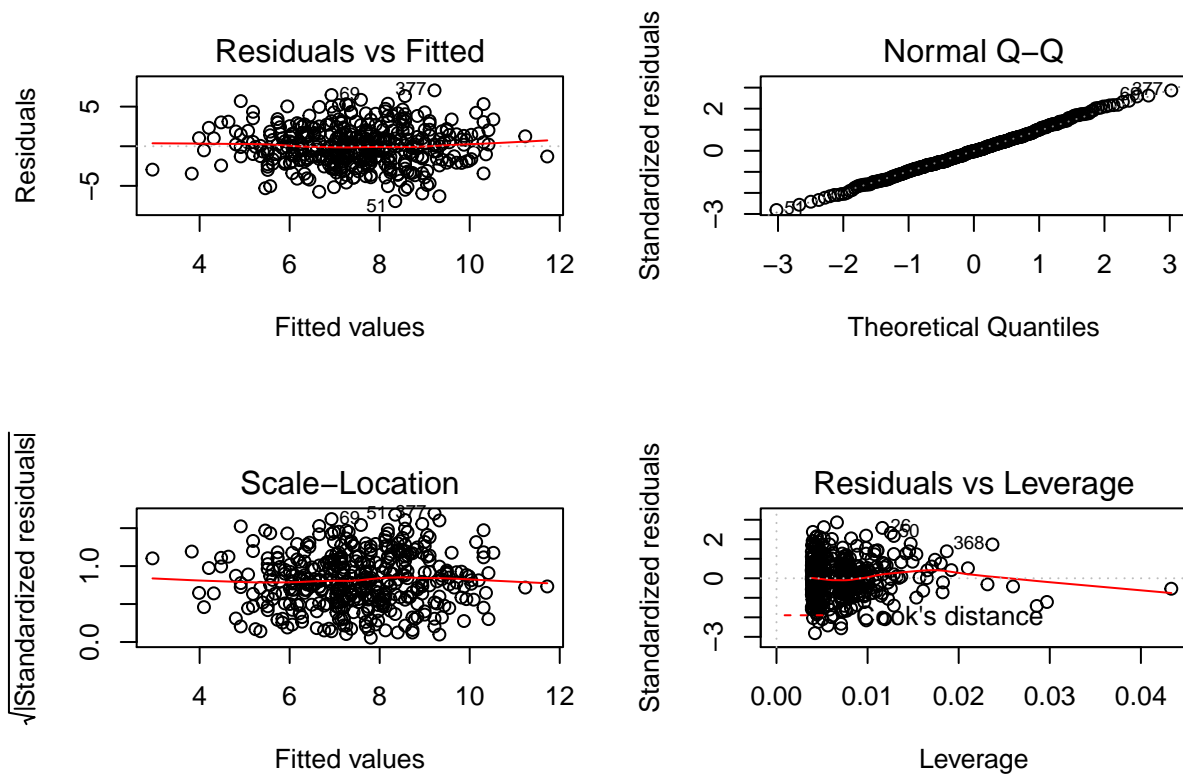
```
# According to the results, price and us are statistically significant.  
#Because the p-values and t-value are smaller than the signifance level,  
#so we can deny the null hypothesis and accept the alternative hypothesis.
```

```
##(e)  
lm.fit2=lm(Sales~Price+US,data=Carseats)  
##the lm.fit2 is the new smaller model, which only uses the  
##significant predictors.
```

```
##(f)  
par(mfrow=c(2,2))  
plot(lm.fit)
```



```
plot(lm.fit2)
```



```
anova(lm.fit,lm.fit2)
```

```
## Analysis of Variance Table
```

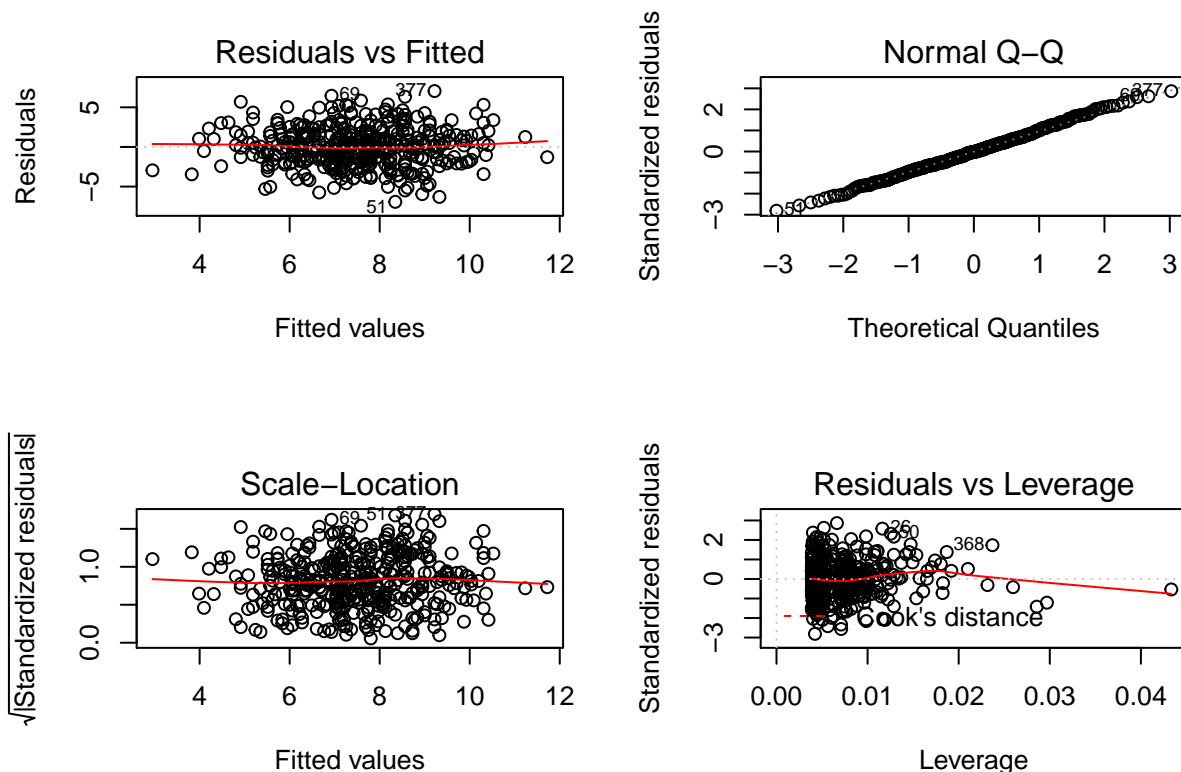
```
##
## Model 1: Sales ~ Price + Urban + US
## Model 2: Sales ~ Price + US
##   Res.Df    RSS Df Sum of Sq    F Pr(>F)
## 1     396 2420.8
## 2     397 2420.9 -1   -0.03979 0.0065 0.9357
```

*#According to the results above, we can conclude that, generally,  
#both of these two models fit well, but have high-leverage and outlier problem  
#according to the plots.  
#And compared between these two models, the difference is small but the  
#latter one might fit better because the R-square is a little bigger, so  
#as anova function shows.*

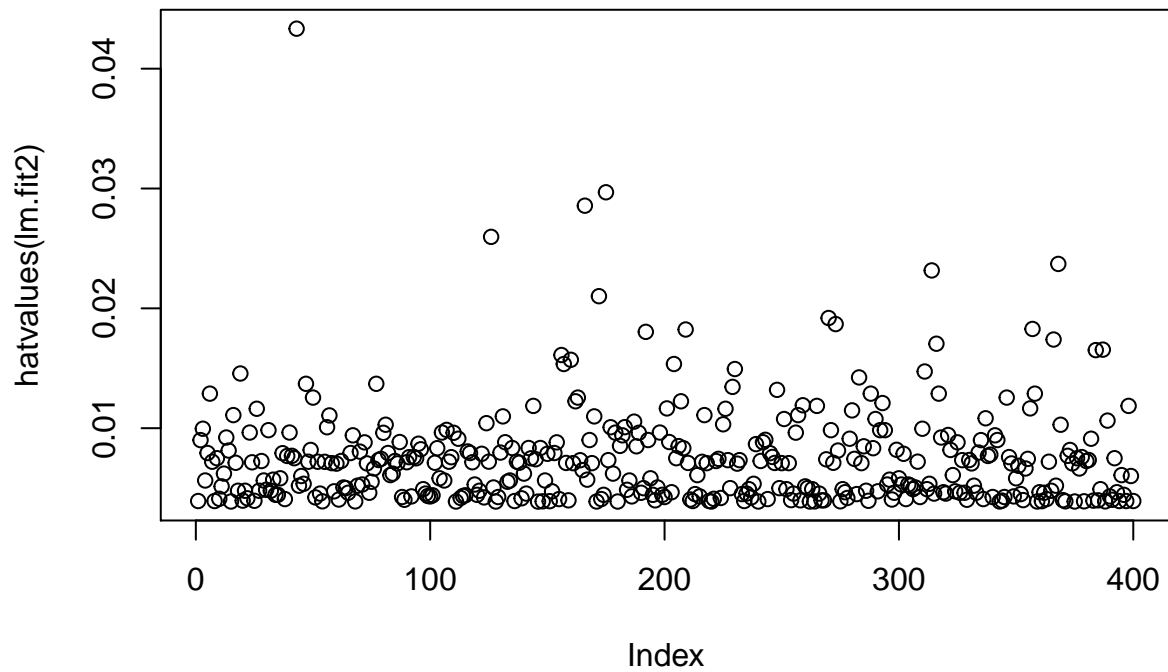
```
##(g)
confint(lm.fit2)
```

```
##              2.5 %      97.5 %
## (Intercept) 11.79032020 14.27126531
## Price       -0.06475984 -0.04419543
## USYes        0.69151957  1.70776632
```

```
##(h)
par(mfrow=c(2,2))
plot(lm.fit2)
```



```
par(mfrow=c(1,1))
plot(hatvalues(lm.fit2))
```



*#According to the plots, we can see that these are outliers and high-leverage  
#problems in the model lm.fit2.  
#In the Residuals vs Leverage plot we can see the high-leverage point 368  
#which has a extremely big x value. Also, in the hatvalue plots, and  
#Residuals vs fitted plots we can see the points with very big Y value,  
#including 69, 377, 51 , are outliers.*