

Worksheet 4c

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1. Use the dataset mpg

1a. Show your solutions on how to import a csv file into the environmen

```
mpg <- read.csv("mpg.csv")
```

1b. Which variables from mpg dataset are categorical?

```
varCat <- mpg[,c(2:4,6:12)]  
varCat
```

##	manufacturer	model	displ	cyl	trans	drv	cty	hwy	fl
## 1	audi	a4	1.8	4	auto(l5)	f	18	29	p
## 2	audi	a4	1.8	4	manual(m5)	f	21	29	p
## 3	audi	a4	2.0	4	manual(m6)	f	20	31	p
## 4	audi	a4	2.0	4	auto(av)	f	21	30	p
## 5	audi	a4	2.8	6	auto(l5)	f	16	26	p
## 6	audi	a4	2.8	6	manual(m5)	f	18	26	p
## 7	audi	a4	3.1	6	auto(av)	f	18	27	p
## 8	audi	a4 quattro	1.8	4	manual(m5)	4	18	26	p
## 9	audi	a4 quattro	1.8	4	auto(l5)	4	16	25	p
## 10	audi	a4 quattro	2.0	4	manual(m6)	4	20	28	p
## 11	audi	a4 quattro	2.0	4	auto(s6)	4	19	27	p
## 12	audi	a4 quattro	2.8	6	auto(l5)	4	15	25	p
## 13	audi	a4 quattro	2.8	6	manual(m5)	4	17	25	p
## 14	audi	a4 quattro	3.1	6	auto(s6)	4	17	25	p
## 15	audi	a4 quattro	3.1	6	manual(m6)	4	15	25	p
## 16	audi	a6 quattro	2.8	6	auto(l5)	4	15	24	p
## 17	audi	a6 quattro	3.1	6	auto(s6)	4	17	25	p
## 18	audi	a6 quattro	4.2	8	auto(s6)	4	16	23	p
## 19	chevrolet	c1500 suburban 2wd	5.3	8	auto(l4)	r	14	20	r
## 20	chevrolet	c1500 suburban 2wd	5.3	8	auto(l4)	r	11	15	e
## 21	chevrolet	c1500 suburban 2wd	5.3	8	auto(l4)	r	14	20	r
## 22	chevrolet	c1500 suburban 2wd	5.7	8	auto(l4)	r	13	17	r
## 23	chevrolet	c1500 suburban 2wd	6.0	8	auto(l4)	r	12	17	r
## 24	chevrolet	corvette	5.7	8	manual(m6)	r	16	26	p
## 25	chevrolet	corvette	5.7	8	auto(l4)	r	15	23	p
## 26	chevrolet	corvette	6.2	8	manual(m6)	r	16	26	p
## 27	chevrolet	corvette	6.2	8	auto(s6)	r	15	25	p
## 28	chevrolet	corvette	7.0	8	manual(m6)	r	15	24	p
## 29	chevrolet	k1500 tahoe 4wd	5.3	8	auto(l4)	4	14	19	r
## 30	chevrolet	k1500 tahoe 4wd	5.3	8	auto(l4)	4	11	14	e
## 31	chevrolet	k1500 tahoe 4wd	5.7	8	auto(l4)	4	11	15	r
## 32	chevrolet	k1500 tahoe 4wd	6.5	8	auto(l4)	4	14	17	d
## 33	chevrolet	malibu	2.4	4	auto(l4)	f	19	27	r

## 34	chevrolet	malibu	2.4	4	auto(14)	f	22	30	r
## 35	chevrolet	malibu	3.1	6	auto(14)	f	18	26	r
## 36	chevrolet	malibu	3.5	6	auto(14)	f	18	29	r
## 37	chevrolet	malibu	3.6	6	auto(s6)	f	17	26	r
## 38	dodge	caravan 2wd	2.4	4	auto(13)	f	18	24	r
## 39	dodge	caravan 2wd	3.0	6	auto(14)	f	17	24	r
## 40	dodge	caravan 2wd	3.3	6	auto(14)	f	16	22	r
## 41	dodge	caravan 2wd	3.3	6	auto(14)	f	16	22	r
## 42	dodge	caravan 2wd	3.3	6	auto(14)	f	17	24	r
## 43	dodge	caravan 2wd	3.3	6	auto(14)	f	17	24	r
## 44	dodge	caravan 2wd	3.3	6	auto(14)	f	11	17	e
## 45	dodge	caravan 2wd	3.8	6	auto(14)	f	15	22	r
## 46	dodge	caravan 2wd	3.8	6	auto(14)	f	15	21	r
## 47	dodge	caravan 2wd	3.8	6	auto(16)	f	16	23	r
## 48	dodge	caravan 2wd	4.0	6	auto(16)	f	16	23	r
## 49	dodge	dakota pickup 4wd	3.7	6	manual(m6)	4	15	19	r
## 50	dodge	dakota pickup 4wd	3.7	6	auto(14)	4	14	18	r
## 51	dodge	dakota pickup 4wd	3.9	6	auto(14)	4	13	17	r
## 52	dodge	dakota pickup 4wd	3.9	6	manual(m5)	4	14	17	r
## 53	dodge	dakota pickup 4wd	4.7	8	auto(15)	4	14	19	r
## 54	dodge	dakota pickup 4wd	4.7	8	auto(15)	4	14	19	r
## 55	dodge	dakota pickup 4wd	4.7	8	auto(15)	4	9	12	e
## 56	dodge	dakota pickup 4wd	5.2	8	manual(m5)	4	11	17	r
## 57	dodge	dakota pickup 4wd	5.2	8	auto(14)	4	11	15	r
## 58	dodge	durango 4wd	3.9	6	auto(14)	4	13	17	r
## 59	dodge	durango 4wd	4.7	8	auto(15)	4	13	17	r
## 60	dodge	durango 4wd	4.7	8	auto(15)	4	9	12	e
## 61	dodge	durango 4wd	4.7	8	auto(15)	4	13	17	r
## 62	dodge	durango 4wd	5.2	8	auto(14)	4	11	16	r
## 63	dodge	durango 4wd	5.7	8	auto(15)	4	13	18	r
## 64	dodge	durango 4wd	5.9	8	auto(14)	4	11	15	r
## 65	dodge	ram 1500 pickup 4wd	4.7	8	manual(m6)	4	12	16	r
## 66	dodge	ram 1500 pickup 4wd	4.7	8	auto(15)	4	9	12	e
## 67	dodge	ram 1500 pickup 4wd	4.7	8	auto(15)	4	13	17	r
## 68	dodge	ram 1500 pickup 4wd	4.7	8	auto(15)	4	13	17	r
## 69	dodge	ram 1500 pickup 4wd	4.7	8	manual(m6)	4	12	16	r
## 70	dodge	ram 1500 pickup 4wd	4.7	8	manual(m6)	4	9	12	e
## 71	dodge	ram 1500 pickup 4wd	5.2	8	auto(14)	4	11	15	r
## 72	dodge	ram 1500 pickup 4wd	5.2	8	manual(m5)	4	11	16	r
## 73	dodge	ram 1500 pickup 4wd	5.7	8	auto(15)	4	13	17	r
## 74	dodge	ram 1500 pickup 4wd	5.9	8	auto(14)	4	11	15	r
## 75	ford	expedition 2wd	4.6	8	auto(14)	r	11	17	r
## 76	ford	expedition 2wd	5.4	8	auto(14)	r	11	17	r
## 77	ford	expedition 2wd	5.4	8	auto(16)	r	12	18	r
## 78	ford	explorer 4wd	4.0	6	auto(15)	4	14	17	r
## 79	ford	explorer 4wd	4.0	6	manual(m5)	4	15	19	r
## 80	ford	explorer 4wd	4.0	6	auto(15)	4	14	17	r
## 81	ford	explorer 4wd	4.0	6	auto(15)	4	13	19	r
## 82	ford	explorer 4wd	4.6	8	auto(16)	4	13	19	r
## 83	ford	explorer 4wd	5.0	8	auto(14)	4	13	17	r
## 84	ford	f150 pickup 4wd	4.2	6	auto(14)	4	14	17	r
## 85	ford	f150 pickup 4wd	4.2	6	manual(m5)	4	14	17	r
## 86	ford	f150 pickup 4wd	4.6	8	manual(m5)	4	13	16	r
## 87	ford	f150 pickup 4wd	4.6	8	auto(14)	4	13	16	r

## 88	ford	f150 pickup 4wd	4.6	8	auto(14)	4	13	17	r
## 89	ford	f150 pickup 4wd	5.4	8	auto(14)	4	11	15	r
## 90	ford	f150 pickup 4wd	5.4	8	auto(14)	4	13	17	r
## 91	ford	mustang	3.8	6	manual(m5)	r	18	26	r
## 92	ford	mustang	3.8	6	auto(14)	r	18	25	r
## 93	ford	mustang	4.0	6	manual(m5)	r	17	26	r
## 94	ford	mustang	4.0	6	auto(15)	r	16	24	r
## 95	ford	mustang	4.6	8	auto(14)	r	15	21	r
## 96	ford	mustang	4.6	8	manual(m5)	r	15	22	r
## 97	ford	mustang	4.6	8	manual(m5)	r	15	23	r
## 98	ford	mustang	4.6	8	auto(15)	r	15	22	r
## 99	ford	mustang	5.4	8	manual(m6)	r	14	20	p
## 100	honda	civic	1.6	4	manual(m5)	f	28	33	r
## 101	honda	civic	1.6	4	auto(14)	f	24	32	r
## 102	honda	civic	1.6	4	manual(m5)	f	25	32	r
## 103	honda	civic	1.6	4	manual(m5)	f	23	29	p
## 104	honda	civic	1.6	4	auto(14)	f	24	32	r
## 105	honda	civic	1.8	4	manual(m5)	f	26	34	r
## 106	honda	civic	1.8	4	auto(15)	f	25	36	r
## 107	honda	civic	1.8	4	auto(15)	f	24	36	c
## 108	honda	civic	2.0	4	manual(m6)	f	21	29	p
## 109	hyundai	sonata	2.4	4	auto(14)	f	18	26	r
## 110	hyundai	sonata	2.4	4	manual(m5)	f	18	27	r
## 111	hyundai	sonata	2.4	4	auto(14)	f	21	30	r
## 112	hyundai	sonata	2.4	4	manual(m5)	f	21	31	r
## 113	hyundai	sonata	2.5	6	auto(14)	f	18	26	r
## 114	hyundai	sonata	2.5	6	manual(m5)	f	18	26	r
## 115	hyundai	sonata	3.3	6	auto(15)	f	19	28	r
## 116	hyundai	tiburon	2.0	4	auto(14)	f	19	26	r
## 117	hyundai	tiburon	2.0	4	manual(m5)	f	19	29	r
## 118	hyundai	tiburon	2.0	4	manual(m5)	f	20	28	r
## 119	hyundai	tiburon	2.0	4	auto(14)	f	20	27	r
## 120	hyundai	tiburon	2.7	6	auto(14)	f	17	24	r
## 121	hyundai	tiburon	2.7	6	manual(m6)	f	16	24	r
## 122	hyundai	tiburon	2.7	6	manual(m5)	f	17	24	r
## 123	jeep	grand cherokee 4wd	3.0	6	auto(15)	4	17	22	d
## 124	jeep	grand cherokee 4wd	3.7	6	auto(15)	4	15	19	r
## 125	jeep	grand cherokee 4wd	4.0	6	auto(14)	4	15	20	r
## 126	jeep	grand cherokee 4wd	4.7	8	auto(14)	4	14	17	r
## 127	jeep	grand cherokee 4wd	4.7	8	auto(15)	4	9	12	e
## 128	jeep	grand cherokee 4wd	4.7	8	auto(15)	4	14	19	r
## 129	jeep	grand cherokee 4wd	5.7	8	auto(15)	4	13	18	r
## 130	jeep	grand cherokee 4wd	6.1	8	auto(15)	4	11	14	p
## 131	land rover	range rover	4.0	8	auto(14)	4	11	15	p
## 132	land rover	range rover	4.2	8	auto(s6)	4	12	18	r
## 133	land rover	range rover	4.4	8	auto(s6)	4	12	18	r
## 134	land rover	range rover	4.6	8	auto(14)	4	11	15	p
## 135	lincoln	navigator 2wd	5.4	8	auto(14)	r	11	17	r
## 136	lincoln	navigator 2wd	5.4	8	auto(14)	r	11	16	p
## 137	lincoln	navigator 2wd	5.4	8	auto(16)	r	12	18	r
## 138	mercury	mountaineer 4wd	4.0	6	auto(15)	4	14	17	r
## 139	mercury	mountaineer 4wd	4.0	6	auto(15)	4	13	19	r
## 140	mercury	mountaineer 4wd	4.6	8	auto(16)	4	13	19	r
## 141	mercury	mountaineer 4wd	5.0	8	auto(14)	4	13	17	r

## 142	nissan	altima	2.4	4	manual(m5)	f	21	29	r
## 143	nissan	altima	2.4	4	auto(14)	f	19	27	r
## 144	nissan	altima	2.5	4	auto(av)	f	23	31	r
## 145	nissan	altima	2.5	4	manual(m6)	f	23	32	r
## 146	nissan	altima	3.5	6	manual(m6)	f	19	27	p
## 147	nissan	altima	3.5	6	auto(av)	f	19	26	p
## 148	nissan	maxima	3.0	6	auto(14)	f	18	26	r
## 149	nissan	maxima	3.0	6	manual(m5)	f	19	25	r
## 150	nissan	maxima	3.5	6	auto(av)	f	19	25	p
## 151	nissan	pathfinder 4wd	3.3	6	auto(14)	4	14	17	r
## 152	nissan	pathfinder 4wd	3.3	6	manual(m5)	4	15	17	r
## 153	nissan	pathfinder 4wd	4.0	6	auto(15)	4	14	20	p
## 154	nissan	pathfinder 4wd	5.6	8	auto(s5)	4	12	18	p
## 155	pontiac	grand prix	3.1	6	auto(14)	f	18	26	r
## 156	pontiac	grand prix	3.8	6	auto(14)	f	16	26	p
## 157	pontiac	grand prix	3.8	6	auto(14)	f	17	27	r
## 158	pontiac	grand prix	3.8	6	auto(14)	f	18	28	r
## 159	pontiac	grand prix	5.3	8	auto(s4)	f	16	25	p
## 160	subaru	forester awd	2.5	4	manual(m5)	4	18	25	r
## 161	subaru	forester awd	2.5	4	auto(14)	4	18	24	r
## 162	subaru	forester awd	2.5	4	manual(m5)	4	20	27	r
## 163	subaru	forester awd	2.5	4	manual(m5)	4	19	25	p
## 164	subaru	forester awd	2.5	4	auto(14)	4	20	26	r
## 165	subaru	forester awd	2.5	4	auto(14)	4	18	23	p
## 166	subaru	impreza awd	2.2	4	auto(14)	4	21	26	r
## 167	subaru	impreza awd	2.2	4	manual(m5)	4	19	26	r
## 168	subaru	impreza awd	2.5	4	manual(m5)	4	19	26	r
## 169	subaru	impreza awd	2.5	4	auto(14)	4	19	26	r
## 170	subaru	impreza awd	2.5	4	auto(s4)	4	20	25	p
## 171	subaru	impreza awd	2.5	4	auto(s4)	4	20	27	r
## 172	subaru	impreza awd	2.5	4	manual(m5)	4	19	25	p
## 173	subaru	impreza awd	2.5	4	manual(m5)	4	20	27	r
## 174	toyota	4runner 4wd	2.7	4	manual(m5)	4	15	20	r
## 175	toyota	4runner 4wd	2.7	4	auto(14)	4	16	20	r
## 176	toyota	4runner 4wd	3.4	6	auto(14)	4	15	19	r
## 177	toyota	4runner 4wd	3.4	6	manual(m5)	4	15	17	r
## 178	toyota	4runner 4wd	4.0	6	auto(15)	4	16	20	r
## 179	toyota	4runner 4wd	4.7	8	auto(15)	4	14	17	r
## 180	toyota	camry	2.2	4	manual(m5)	f	21	29	r
## 181	toyota	camry	2.2	4	auto(14)	f	21	27	r
## 182	toyota	camry	2.4	4	manual(m5)	f	21	31	r
## 183	toyota	camry	2.4	4	auto(15)	f	21	31	r
## 184	toyota	camry	3.0	6	auto(14)	f	18	26	r
## 185	toyota	camry	3.0	6	manual(m5)	f	18	26	r
## 186	toyota	camry	3.5	6	auto(s6)	f	19	28	r
## 187	toyota	camry solara	2.2	4	auto(14)	f	21	27	r
## 188	toyota	camry solara	2.2	4	manual(m5)	f	21	29	r
## 189	toyota	camry solara	2.4	4	manual(m5)	f	21	31	r
## 190	toyota	camry solara	2.4	4	auto(s5)	f	22	31	r
## 191	toyota	camry solara	3.0	6	auto(14)	f	18	26	r
## 192	toyota	camry solara	3.0	6	manual(m5)	f	18	26	r
## 193	toyota	camry solara	3.3	6	auto(s5)	f	18	27	r
## 194	toyota	corolla	1.8	4	auto(13)	f	24	30	r
## 195	toyota	corolla	1.8	4	auto(14)	f	24	33	r

## 196	toyota	corolla	1.8	4 manual(m5)	f	26	35	r
## 197	toyota	corolla	1.8	4 manual(m5)	f	28	37	r
## 198	toyota	corolla	1.8	4 auto(14)	f	26	35	r
## 199	toyota	land cruiser wagon 4wd	4.7	8 auto(14)	4	11	15	r
## 200	toyota	land cruiser wagon 4wd	5.7	8 auto(s6)	4	13	18	r
## 201	toyota	toyota tacoma 4wd	2.7	4 manual(m5)	4	15	20	r
## 202	toyota	toyota tacoma 4wd	2.7	4 auto(14)	4	16	20	r
## 203	toyota	toyota tacoma 4wd	2.7	4 manual(m5)	4	17	22	r
## 204	toyota	toyota tacoma 4wd	3.4	6 manual(m5)	4	15	17	r
## 205	toyota	toyota tacoma 4wd	3.4	6 auto(14)	4	15	19	r
## 206	toyota	toyota tacoma 4wd	4.0	6 manual(m6)	4	15	18	r
## 207	toyota	toyota tacoma 4wd	4.0	6 auto(15)	4	16	20	r
## 208	volkswagen	gti	2.0	4 manual(m5)	f	21	29	r
## 209	volkswagen	gti	2.0	4 auto(14)	f	19	26	r
## 210	volkswagen	gti	2.0	4 manual(m6)	f	21	29	p
## 211	volkswagen	gti	2.0	4 auto(s6)	f	22	29	p
## 212	volkswagen	gti	2.8	6 manual(m5)	f	17	24	r
## 213	volkswagen	jetta	1.9	4 manual(m5)	f	33	44	d
## 214	volkswagen	jetta	2.0	4 manual(m5)	f	21	29	r
## 215	volkswagen	jetta	2.0	4 auto(14)	f	19	26	r
## 216	volkswagen	jetta	2.0	4 auto(s6)	f	22	29	p
## 217	volkswagen	jetta	2.0	4 manual(m6)	f	21	29	p
## 218	volkswagen	jetta	2.5	5 auto(s6)	f	21	29	r
## 219	volkswagen	jetta	2.5	5 manual(m5)	f	21	29	r
## 220	volkswagen	jetta	2.8	6 auto(14)	f	16	23	r
## 221	volkswagen	jetta	2.8	6 manual(m5)	f	17	24	r
## 222	volkswagen	new beetle	1.9	4 manual(m5)	f	35	44	d
## 223	volkswagen	new beetle	1.9	4 auto(14)	f	29	41	d
## 224	volkswagen	new beetle	2.0	4 manual(m5)	f	21	29	r
## 225	volkswagen	new beetle	2.0	4 auto(14)	f	19	26	r
## 226	volkswagen	new beetle	2.5	5 manual(m5)	f	20	28	r
## 227	volkswagen	new beetle	2.5	5 auto(s6)	f	20	29	r
## 228	volkswagen	passat	1.8	4 manual(m5)	f	21	29	p
## 229	volkswagen	passat	1.8	4 auto(15)	f	18	29	p
## 230	volkswagen	passat	2.0	4 auto(s6)	f	19	28	p
## 231	volkswagen	passat	2.0	4 manual(m6)	f	21	29	p
## 232	volkswagen	passat	2.8	6 auto(15)	f	16	26	p
## 233	volkswagen	passat	2.8	6 manual(m5)	f	18	26	p
## 234	volkswagen	passat	3.6	6 auto(s6)	f	17	26	p
##	class							
## 1	compact							
## 2	compact							
## 3	compact							
## 4	compact							
## 5	compact							
## 6	compact							
## 7	compact							
## 8	compact							
## 9	compact							
## 10	compact							
## 11	compact							
## 12	compact							
## 13	compact							
## 14	compact							

## 15	compact
## 16	midsize
## 17	midsize
## 18	midsize
## 19	suv
## 20	suv
## 21	suv
## 22	suv
## 23	suv
## 24	2seater
## 25	2seater
## 26	2seater
## 27	2seater
## 28	2seater
## 29	suv
## 30	suv
## 31	suv
## 32	suv
## 33	midsize
## 34	midsize
## 35	midsize
## 36	midsize
## 37	midsize
## 38	minivan
## 39	minivan
## 40	minivan
## 41	minivan
## 42	minivan
## 43	minivan
## 44	minivan
## 45	minivan
## 46	minivan
## 47	minivan
## 48	minivan
## 49	pickup
## 50	pickup
## 51	pickup
## 52	pickup
## 53	pickup
## 54	pickup
## 55	pickup
## 56	pickup
## 57	pickup
## 58	suv
## 59	suv
## 60	suv
## 61	suv
## 62	suv
## 63	suv
## 64	suv
## 65	pickup
## 66	pickup
## 67	pickup
## 68	pickup

69 pickup
70 pickup
71 pickup
72 pickup
73 pickup
74 pickup
75 suv
76 suv
77 suv
78 suv
79 suv
80 suv
81 suv
82 suv
83 suv
84 pickup
85 pickup
86 pickup
87 pickup
88 pickup
89 pickup
90 pickup
91 subcompact
92 subcompact
93 subcompact
94 subcompact
95 subcompact
96 subcompact
97 subcompact
98 subcompact
99 subcompact
100 subcompact
101 subcompact
102 subcompact
103 subcompact
104 subcompact
105 subcompact
106 subcompact
107 subcompact
108 subcompact
109 midsize
110 midsize
111 midsize
112 midsize
113 midsize
114 midsize
115 midsize
116 subcompact
117 subcompact
118 subcompact
119 subcompact
120 subcompact
121 subcompact
122 subcompact

## 123	suv
## 124	suv
## 125	suv
## 126	suv
## 127	suv
## 128	suv
## 129	suv
## 130	suv
## 131	suv
## 132	suv
## 133	suv
## 134	suv
## 135	suv
## 136	suv
## 137	suv
## 138	suv
## 139	suv
## 140	suv
## 141	suv
## 142	compact
## 143	compact
## 144	midsize
## 145	midsize
## 146	midsize
## 147	midsize
## 148	midsize
## 149	midsize
## 150	midsize
## 151	suv
## 152	suv
## 153	suv
## 154	suv
## 155	midsize
## 156	midsize
## 157	midsize
## 158	midsize
## 159	midsize
## 160	suv
## 161	suv
## 162	suv
## 163	suv
## 164	suv
## 165	suv
## 166	subcompact
## 167	subcompact
## 168	subcompact
## 169	subcompact
## 170	compact
## 171	compact
## 172	compact
## 173	compact
## 174	suv
## 175	suv
## 176	suv

## 177	suv
## 178	suv
## 179	suv
## 180	midsize
## 181	midsize
## 182	midsize
## 183	midsize
## 184	midsize
## 185	midsize
## 186	midsize
## 187	compact
## 188	compact
## 189	compact
## 190	compact
## 191	compact
## 192	compact
## 193	compact
## 194	compact
## 195	compact
## 196	compact
## 197	compact
## 198	compact
## 199	suv
## 200	suv
## 201	pickup
## 202	pickup
## 203	pickup
## 204	pickup
## 205	pickup
## 206	pickup
## 207	pickup
## 208	compact
## 209	compact
## 210	compact
## 211	compact
## 212	compact
## 213	compact
## 214	compact
## 215	compact
## 216	compact
## 217	compact
## 218	compact
## 219	compact
## 220	compact
## 221	compact
## 222	subcompact
## 223	subcompact
## 224	subcompact
## 225	subcompact
## 226	subcompact
## 227	subcompact
## 228	midsize
## 229	midsize
## 230	midsize

```
## 231    midsize
## 232    midsize
## 233    midsize
## 234    midsize
```

1c. Which are continuous variables?

```
carCon <- mpg[,c(4:6,9,10)]
carCon
```

```
##      displ year  cyl  cty hwy
## 1      1.8 1999   4   18  29
## 2      1.8 1999   4   21  29
## 3      2.0 2008   4   20  31
## 4      2.0 2008   4   21  30
## 5      2.8 1999   6   16  26
## 6      2.8 1999   6   18  26
## 7      3.1 2008   6   18  27
## 8      1.8 1999   4   18  26
## 9      1.8 1999   4   16  25
## 10     2.0 2008   4   20  28
## 11     2.0 2008   4   19  27
## 12     2.8 1999   6   15  25
## 13     2.8 1999   6   17  25
## 14     3.1 2008   6   17  25
## 15     3.1 2008   6   15  25
## 16     2.8 1999   6   15  24
## 17     3.1 2008   6   17  25
## 18     4.2 2008   8   16  23
## 19     5.3 2008   8   14  20
## 20     5.3 2008   8   11  15
## 21     5.3 2008   8   14  20
## 22     5.7 1999   8   13  17
## 23     6.0 2008   8   12  17
## 24     5.7 1999   8   16  26
## 25     5.7 1999   8   15  23
## 26     6.2 2008   8   16  26
## 27     6.2 2008   8   15  25
## 28     7.0 2008   8   15  24
## 29     5.3 2008   8   14  19
## 30     5.3 2008   8   11  14
## 31     5.7 1999   8   11  15
## 32     6.5 1999   8   14  17
## 33     2.4 1999   4   19  27
## 34     2.4 2008   4   22  30
## 35     3.1 1999   6   18  26
## 36     3.5 2008   6   18  29
## 37     3.6 2008   6   17  26
## 38     2.4 1999   4   18  24
## 39     3.0 1999   6   17  24
## 40     3.3 1999   6   16  22
## 41     3.3 1999   6   16  22
## 42     3.3 2008   6   17  24
## 43     3.3 2008   6   17  24
## 44     3.3 2008   6   11  17
```

## 45	3.8	1999	6	15	22
## 46	3.8	1999	6	15	21
## 47	3.8	2008	6	16	23
## 48	4.0	2008	6	16	23
## 49	3.7	2008	6	15	19
## 50	3.7	2008	6	14	18
## 51	3.9	1999	6	13	17
## 52	3.9	1999	6	14	17
## 53	4.7	2008	8	14	19
## 54	4.7	2008	8	14	19
## 55	4.7	2008	8	9	12
## 56	5.2	1999	8	11	17
## 57	5.2	1999	8	11	15
## 58	3.9	1999	6	13	17
## 59	4.7	2008	8	13	17
## 60	4.7	2008	8	9	12
## 61	4.7	2008	8	13	17
## 62	5.2	1999	8	11	16
## 63	5.7	2008	8	13	18
## 64	5.9	1999	8	11	15
## 65	4.7	2008	8	12	16
## 66	4.7	2008	8	9	12
## 67	4.7	2008	8	13	17
## 68	4.7	2008	8	13	17
## 69	4.7	2008	8	12	16
## 70	4.7	2008	8	9	12
## 71	5.2	1999	8	11	15
## 72	5.2	1999	8	11	16
## 73	5.7	2008	8	13	17
## 74	5.9	1999	8	11	15
## 75	4.6	1999	8	11	17
## 76	5.4	1999	8	11	17
## 77	5.4	2008	8	12	18
## 78	4.0	1999	6	14	17
## 79	4.0	1999	6	15	19
## 80	4.0	1999	6	14	17
## 81	4.0	2008	6	13	19
## 82	4.6	2008	8	13	19
## 83	5.0	1999	8	13	17
## 84	4.2	1999	6	14	17
## 85	4.2	1999	6	14	17
## 86	4.6	1999	8	13	16
## 87	4.6	1999	8	13	16
## 88	4.6	2008	8	13	17
## 89	5.4	1999	8	11	15
## 90	5.4	2008	8	13	17
## 91	3.8	1999	6	18	26
## 92	3.8	1999	6	18	25
## 93	4.0	2008	6	17	26
## 94	4.0	2008	6	16	24
## 95	4.6	1999	8	15	21
## 96	4.6	1999	8	15	22
## 97	4.6	2008	8	15	23
## 98	4.6	2008	8	15	22

## 99	5.4	2008	8	14	20
## 100	1.6	1999	4	28	33
## 101	1.6	1999	4	24	32
## 102	1.6	1999	4	25	32
## 103	1.6	1999	4	23	29
## 104	1.6	1999	4	24	32
## 105	1.8	2008	4	26	34
## 106	1.8	2008	4	25	36
## 107	1.8	2008	4	24	36
## 108	2.0	2008	4	21	29
## 109	2.4	1999	4	18	26
## 110	2.4	1999	4	18	27
## 111	2.4	2008	4	21	30
## 112	2.4	2008	4	21	31
## 113	2.5	1999	6	18	26
## 114	2.5	1999	6	18	26
## 115	3.3	2008	6	19	28
## 116	2.0	1999	4	19	26
## 117	2.0	1999	4	19	29
## 118	2.0	2008	4	20	28
## 119	2.0	2008	4	20	27
## 120	2.7	2008	6	17	24
## 121	2.7	2008	6	16	24
## 122	2.7	2008	6	17	24
## 123	3.0	2008	6	17	22
## 124	3.7	2008	6	15	19
## 125	4.0	1999	6	15	20
## 126	4.7	1999	8	14	17
## 127	4.7	2008	8	9	12
## 128	4.7	2008	8	14	19
## 129	5.7	2008	8	13	18
## 130	6.1	2008	8	11	14
## 131	4.0	1999	8	11	15
## 132	4.2	2008	8	12	18
## 133	4.4	2008	8	12	18
## 134	4.6	1999	8	11	15
## 135	5.4	1999	8	11	17
## 136	5.4	1999	8	11	16
## 137	5.4	2008	8	12	18
## 138	4.0	1999	6	14	17
## 139	4.0	2008	6	13	19
## 140	4.6	2008	8	13	19
## 141	5.0	1999	8	13	17
## 142	2.4	1999	4	21	29
## 143	2.4	1999	4	19	27
## 144	2.5	2008	4	23	31
## 145	2.5	2008	4	23	32
## 146	3.5	2008	6	19	27
## 147	3.5	2008	6	19	26
## 148	3.0	1999	6	18	26
## 149	3.0	1999	6	19	25
## 150	3.5	2008	6	19	25
## 151	3.3	1999	6	14	17
## 152	3.3	1999	6	15	17

## 153	4.0	2008	6	14	20
## 154	5.6	2008	8	12	18
## 155	3.1	1999	6	18	26
## 156	3.8	1999	6	16	26
## 157	3.8	1999	6	17	27
## 158	3.8	2008	6	18	28
## 159	5.3	2008	8	16	25
## 160	2.5	1999	4	18	25
## 161	2.5	1999	4	18	24
## 162	2.5	2008	4	20	27
## 163	2.5	2008	4	19	25
## 164	2.5	2008	4	20	26
## 165	2.5	2008	4	18	23
## 166	2.2	1999	4	21	26
## 167	2.2	1999	4	19	26
## 168	2.5	1999	4	19	26
## 169	2.5	1999	4	19	26
## 170	2.5	2008	4	20	25
## 171	2.5	2008	4	20	27
## 172	2.5	2008	4	19	25
## 173	2.5	2008	4	20	27
## 174	2.7	1999	4	15	20
## 175	2.7	1999	4	16	20
## 176	3.4	1999	6	15	19
## 177	3.4	1999	6	15	17
## 178	4.0	2008	6	16	20
## 179	4.7	2008	8	14	17
## 180	2.2	1999	4	21	29
## 181	2.2	1999	4	21	27
## 182	2.4	2008	4	21	31
## 183	2.4	2008	4	21	31
## 184	3.0	1999	6	18	26
## 185	3.0	1999	6	18	26
## 186	3.5	2008	6	19	28
## 187	2.2	1999	4	21	27
## 188	2.2	1999	4	21	29
## 189	2.4	2008	4	21	31
## 190	2.4	2008	4	22	31
## 191	3.0	1999	6	18	26
## 192	3.0	1999	6	18	26
## 193	3.3	2008	6	18	27
## 194	1.8	1999	4	24	30
## 195	1.8	1999	4	24	33
## 196	1.8	1999	4	26	35
## 197	1.8	2008	4	28	37
## 198	1.8	2008	4	26	35
## 199	4.7	1999	8	11	15
## 200	5.7	2008	8	13	18
## 201	2.7	1999	4	15	20
## 202	2.7	1999	4	16	20
## 203	2.7	2008	4	17	22
## 204	3.4	1999	6	15	17
## 205	3.4	1999	6	15	19
## 206	4.0	2008	6	15	18

```
## 207 4.0 2008 6 16 20
## 208 2.0 1999 4 21 29
## 209 2.0 1999 4 19 26
## 210 2.0 2008 4 21 29
## 211 2.0 2008 4 22 29
## 212 2.8 1999 6 17 24
## 213 1.9 1999 4 33 44
## 214 2.0 1999 4 21 29
## 215 2.0 1999 4 19 26
## 216 2.0 2008 4 22 29
## 217 2.0 2008 4 21 29
## 218 2.5 2008 5 21 29
## 219 2.5 2008 5 21 29
## 220 2.8 1999 6 16 23
## 221 2.8 1999 6 17 24
## 222 1.9 1999 4 35 44
## 223 1.9 1999 4 29 41
## 224 2.0 1999 4 21 29
## 225 2.0 1999 4 19 26
## 226 2.5 2008 5 20 28
## 227 2.5 2008 5 20 29
## 228 1.8 1999 4 21 29
## 229 1.8 1999 4 18 29
## 230 2.0 2008 4 19 28
## 231 2.0 2008 4 21 29
## 232 2.8 1999 6 16 26
## 233 2.8 1999 6 18 26
## 234 3.6 2008 6 17 26
```

2. Which manufacturer has the most models in this data set? Which model has the most variations? Show your answer.

```
maxmod <- table(mpg$manufacturer)
maxmod
```

```
##
##      audi  chevrolet      dodge      ford      honda  hyundai  jeep
##      18      19      37      25      9      14      8
## land rover  lincoln  mercury  nissan  pontiac  subaru  toyota
##      4      3      4      13      5      14      34
## volkswagen
##      27
```

```
#dodge has most model
```

```
maxvarofmod <- table(mpg$model)
maxvarofmod
```

```
##
##      4runner 4wd      a4      a4 quattro
##      6      7      8
##      a6 quattro      altima  c1500 suburban 2wd
##      3      6      5
##      camry      camry solara  caravan 2wd
##      7      7      11
```

```
##           civic           corolla           corvette
##           9             5             5
##   dakota pickup 4wd      durango 4wd      expedition 2wd
##           9             7             3
##   explorer 4wd      f150 pickup 4wd      forester awd
##           6             7             6
##   grand cherokee 4wd      grand prix      gti
##           8             5             5
##   impreza awd           jetta      k1500 tahoe 4wd
##           8             9             4
##   land cruiser wagon 4wd      malibu      maxima
##           2             5             3
##   mountaineer 4wd      mustang      navigator 2wd
##           4             9             3
##   new beetle           passat      pathfinder 4wd
##           6             7             4
##   ram 1500 pickup 4wd      range rover      sonata
##           10            4             7
##   tiburon      toyota tacoma 4wd
##           7             7
```

#caravan 2wd has the most Variations

2a. Group the manufacturers and find the unique models. Show your codes and result.

```
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
```

```
manufacturer_groups <- mpg %>%
  group_by(mpg$manufacturer) %>%
  summarise(unique_models = toString(unique(model)))
manufacturer_groups
```

```
## # A tibble: 15 x 2
##   `mpg$manufacturer` unique_models
##   <chr>              <chr>
## 1 audi              a4, a4 quattro, a6 quattro
## 2 chevrolet         c1500 suburban 2wd, corvette, k1500 tahoe 4wd, malibu
## 3 dodge             caravan 2wd, dakota pickup 4wd, durango 4wd, ram 1500 pic~
## 4 ford             expedition 2wd, explorer 4wd, f150 pickup 4wd, mustang
## 5 honda            civic
## 6 hyundai          sonata, tiburon
## 7 jeep             grand cherokee 4wd
## 8 land rover       range rover
## 9 lincoln          navigator 2wd
## 10 mercury         mountaineer 4wd
## 11 nissan           altima, maxima, pathfinder 4wd
```

```
## 12 pontiac          grand prix
## 13 subaru           forester awd, impreza awd
## 14 toyota           4runner 4wd, camry, camry solara, corolla, land cruiser w~
## 15 volkswagen       gti, jetta, new beetle, passat
```

```
manufacturer_counts <- mpg %>%
  group_by(manufacturer) %>%
  summarise(unique_model_count = n_distinct(model))
manufacturer_counts
```

```
## # A tibble: 15 x 2
##   manufacturer unique_model_count
##   <chr>          <int>
## 1 audi           3
## 2 chevrolet      4
## 3 dodge          4
## 4 ford           4
## 5 honda          1
## 6 hyundai        2
## 7 jeep           1
## 8 land rover     1
## 9 lincoln        1
## 10 mercury        1
## 11 nissan          3
## 12 pontiac        1
## 13 subaru         2
## 14 toyota         6
## 15 volkswagen     4
```

2b. Graph the result by using plot() and ggplot(). Write the codes and its result.

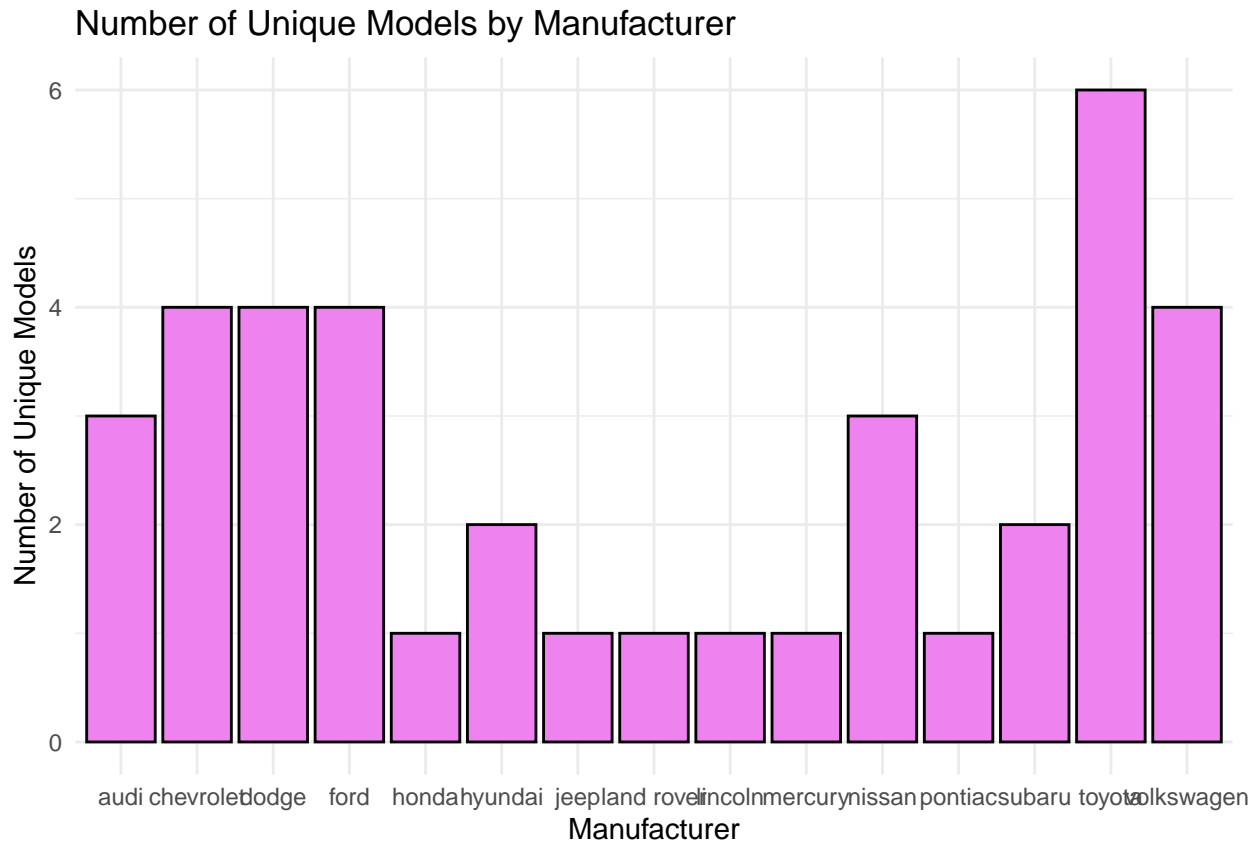
```
par(mfrow = c(1,2))

barplot(manufacturer_counts$unique_model_count, names.arg = manufacturer_counts$manufacturer,
        col = "violet", border = "black", main = "Number of Unique Models by Manufacturer",
        xlab = "Manufacturer", ylab = "Number of Unique Models",
        las = 2)
```

```
library(ggplot2)
```

```
##
## Attaching package: 'ggplot2'
##
## The following object is masked _by_ '.GlobalEnv':
##
##   mpg
ggplotofUMM <- ggplot(manufacturer_counts, aes(x = manufacturer, y = unique_model_count)) +
  geom_bar(stat = "identity", fill = "violet", color = "black") +
  labs(title = "Number of Unique Models by Manufacturer",
       x = "Manufacturer",
       y = "Number of Unique Models") +
  theme_minimal()
ggplotofUMM

par(mfrow = c(1,1))
```

3. Same dataset will be used. You are going to show the relationship of the model and the manufacturer.

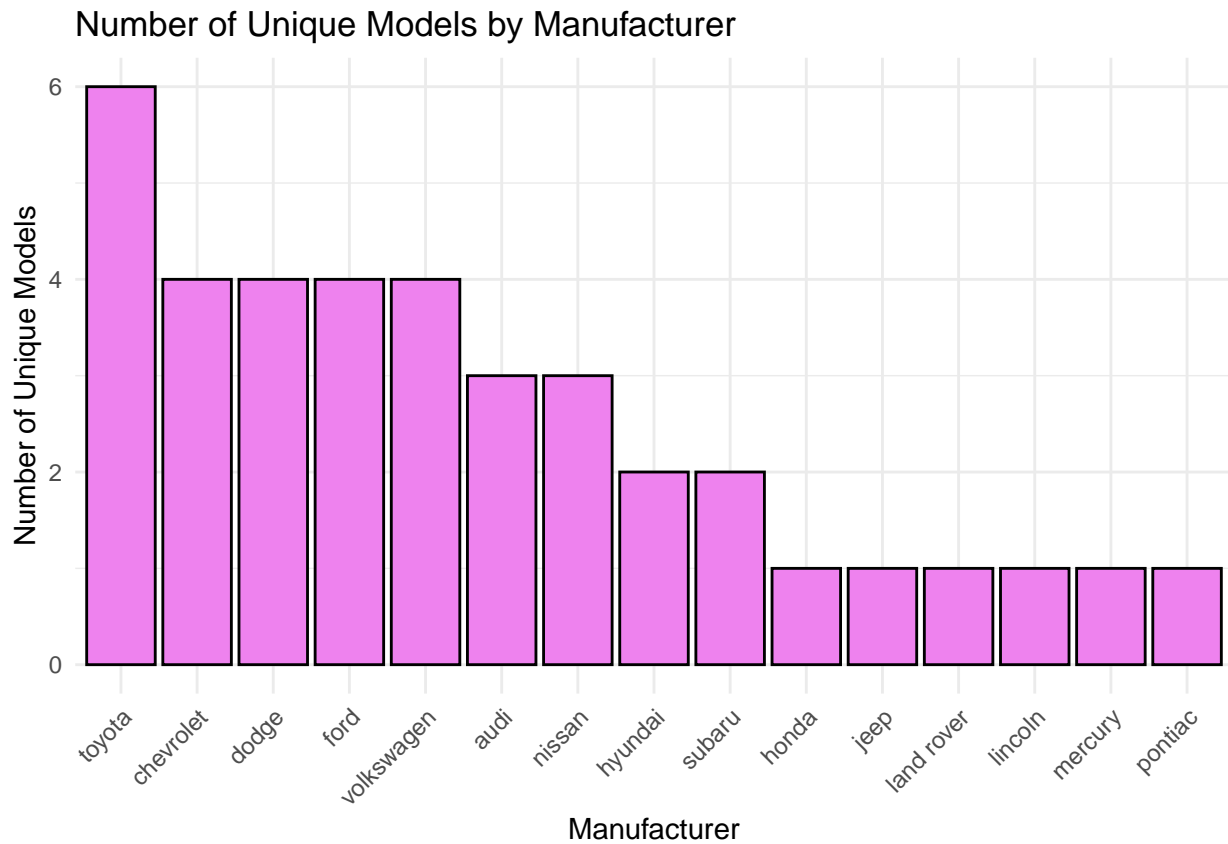
```
manufacturer_model_counts <- mpg %>%
  group_by(mpg$manufacturer, mpg$model) %>%
  summarise(count = n(), .groups = "drop")
```

```
manufacturer_model_counts
```

```
## # A tibble: 38 x 3
##   `mpg$manufacturer` `mpg$model`      count
##   <chr>              <chr>          <int>
## 1 audi              a4              7
## 2 audi              a4 quattro      8
## 3 audi              a6 quattro      3
## 4 chevrolet         c1500 suburban 2wd  5
## 5 chevrolet         corvette          5
## 6 chevrolet         k1500 tahoe 4wd   4
## 7 chevrolet         malibu            5
## 8 dodge             caravan 2wd     11
## 9 dodge             dakota pickup 4wd  9
## 10 dodge            durango 4wd       7
## # i 28 more rows
```

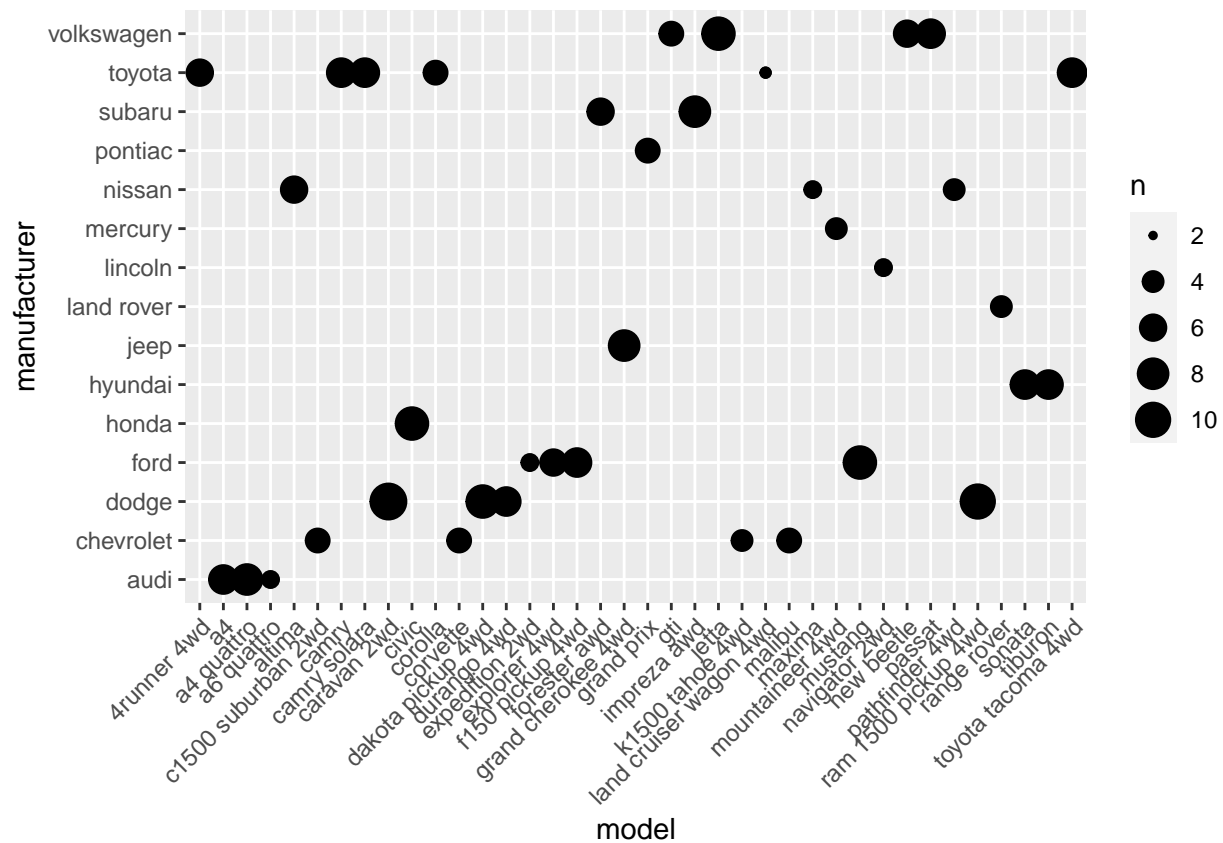
```
ggplot(manufacturer_counts, aes(x = reorder(manufacturer, -unique_model_count), y = unique_model_count)) +
  geom_bar(stat = "identity", fill = "violet", color = "black") +
  labs(title = "Number of Unique Models by Manufacturer",
       x = "Manufacturer",
       y = "Number of Unique Models") +
```

```
theme_minimal() +
theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



3a. What does `ggplot(mpg, aes(model, manufacturer)) + geom_point()` show?

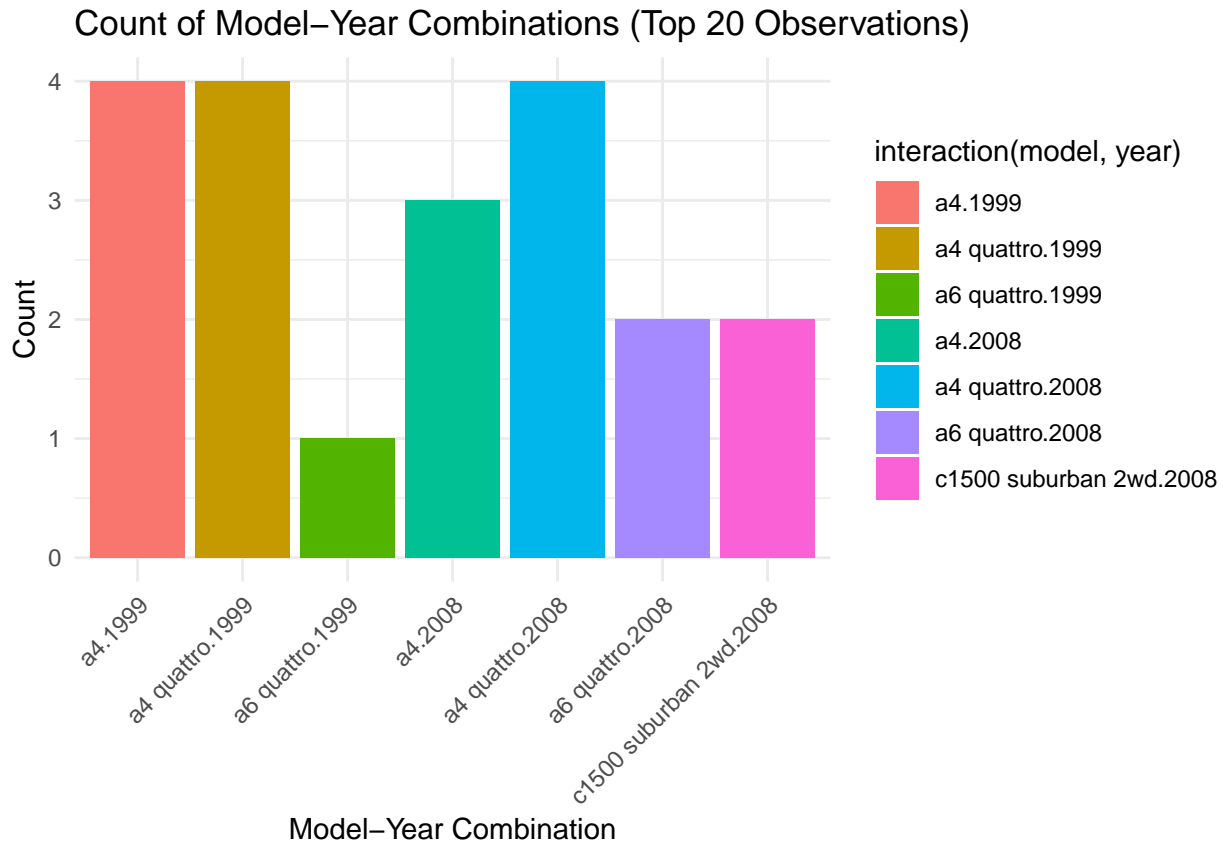
```
ggplot(mpg, aes(model, manufacturer)) +
  geom_point() +
  geom_count() +
  theme(plot.title = element_text(hjust = 0.5),
        axis.text.x = element_text(angle = 45, hjust = 1))
```



3b. For you, is it useful? If not, how could you modify the data to make it more informative?

not really. I would add color

```
ggplot(mpg, aes(model, manufacturer, color = factor(cyl))) +
  geom_point() +
  geom_count() +
  theme(plot.title = element_text(hjust = 0.5),
        axis.text.x = element_text(angle = 45, hjust = 1))
```

4. Using the pipe (`%>%`), group the model and get the number of cars per model. Show codes and its result

```
carsPerModel <- mpg %>%
  group_by(model) %>%
  summarise(number_of_cars = n())
```

carsPerModel

```
## # A tibble: 38 x 2
##   model          number_of_cars
##   <chr>              <int>
## 1 4runner 4wd             6
## 2 a4                     7
## 3 a4 quattro             8
## 4 a6 quattro             3
## 5 altima                 6
## 6 c1500 suburban 2wd     5
## 7 camry                  7
## 8 camry solara           7
## 9 caravan 2wd           11
## 10 civic                  9
## # i 28 more rows
```

- 4a. Plot using `geom_bar()` using the top 20 observations only. The graphs should have a title, labels and colors. Show code and results.

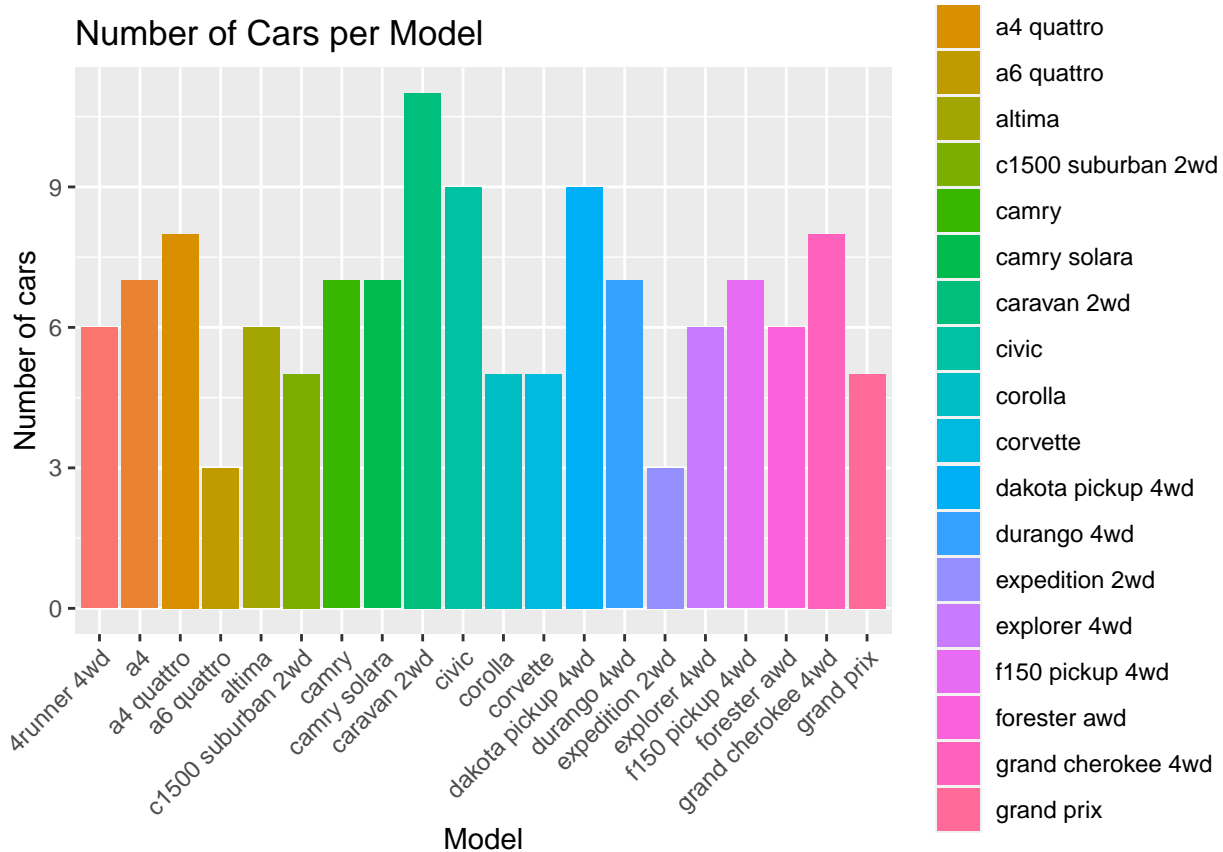
```
top20 <- head(carsPerModel, 20)
```

```

plotofTop20 <- ggplot(top20, aes(x = model, y = number_of_cars, fill = model)) +
  geom_bar(stat = "identity") +
  labs(title = "Number of Cars per Model",
       x = "Model" ,
       y = "Number of cars") + theme(axis.text.x = element_text(angle = 45, hjust = 1))

```

plotofTop20



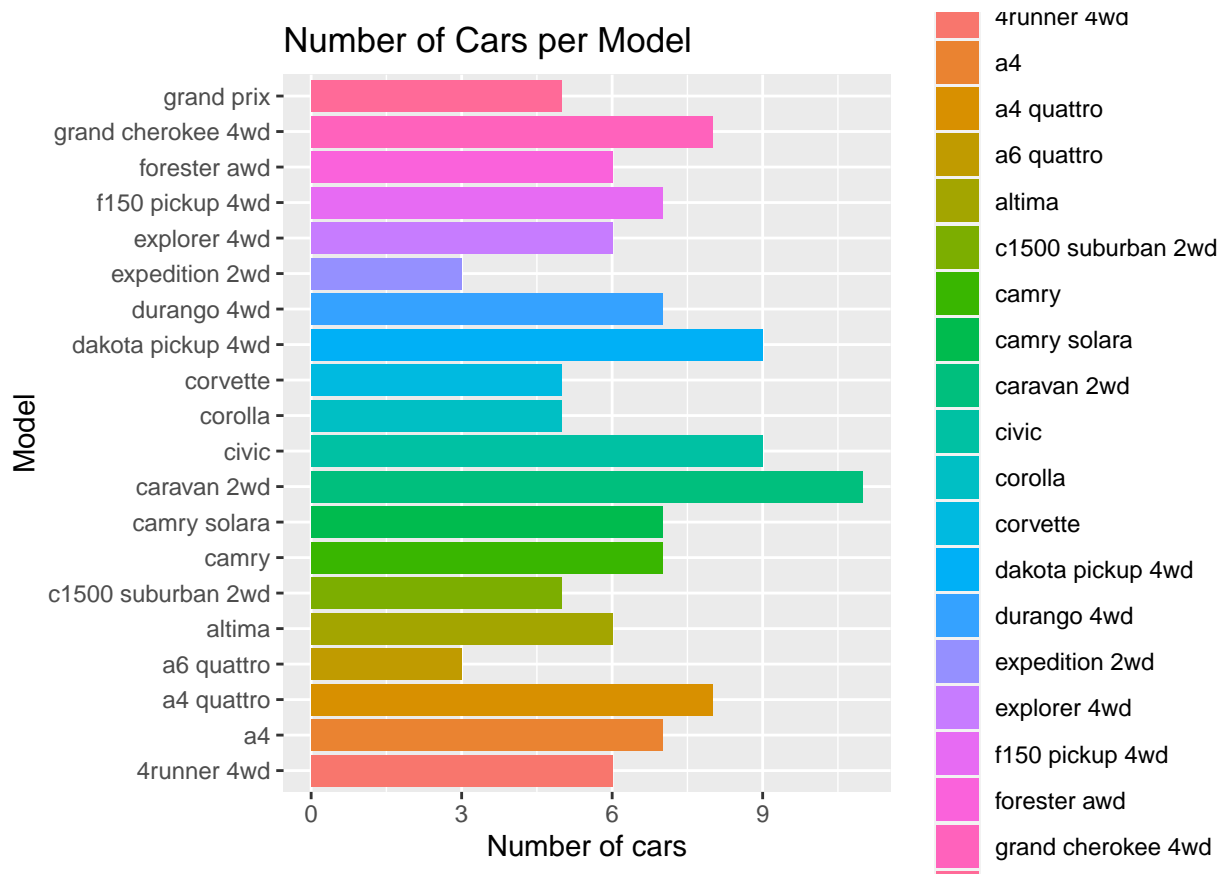
4b. Plot using the `geom_bar()` + `coord_flip()` just like what is shown below. Show codes and its result.

```

lotofTop20 <- ggplot(top20, aes(x = model, y = number_of_cars, fill = model)) +
  geom_bar(stat = "identity") +
  labs(title = "Number of Cars per Model",
       x = "Model" ,
       y = "Number of cars") +
  coord_flip()

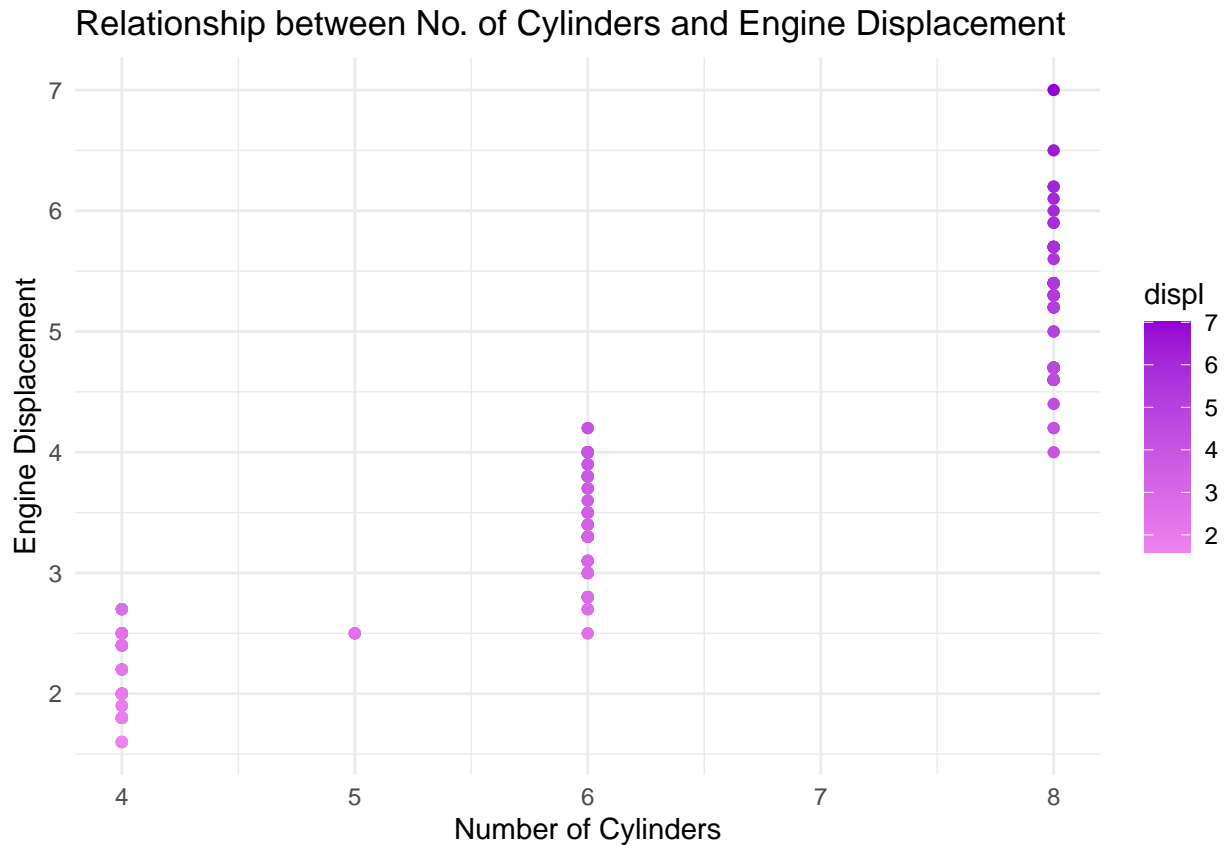
```

`print(lotofTop20)`



5. Plot the relationship between cyl - number of cylinders and displ - engine displacement using `geom_point` with aesthetic color = engine displacement. Title should be "Relationship between No. of Cylinders and Engine Displacement".

```
ggplot(mpg, aes(x = cyl, y = displ, color = displ)) +
  geom_point() +
  scale_color_gradient(low = "violet", high = "darkviolet") +
  labs(title = "Relationship between No. of Cylinders and Engine Displacement",
       x = "Number of Cylinders",
       y = "Engine Displacement") +
  theme_minimal()
```



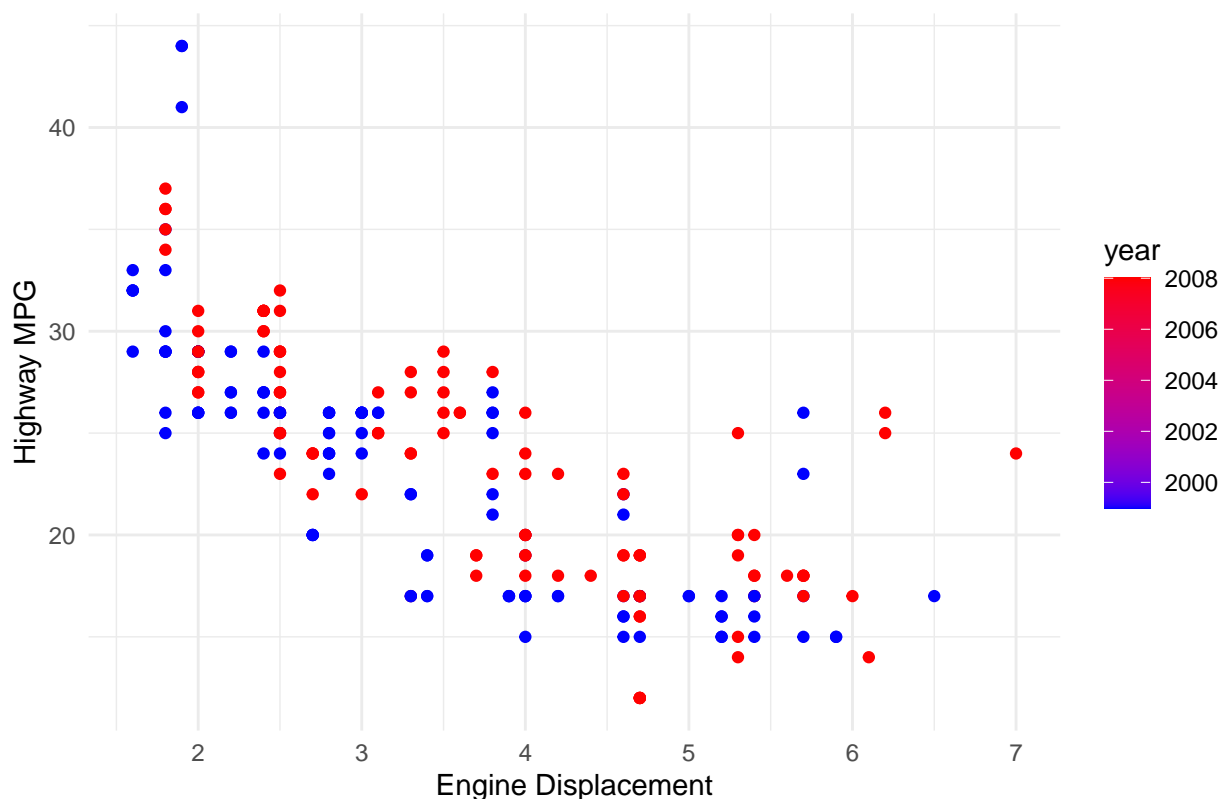
5a. How would you describe its relationship? Show the codes and its result

```
# number of cylenders increases the displacement also increases
```

6. Plot the relationship between displ (engine displacement) and hwy(highway miles per gallon). Mapped it with a continuous variable you have identified in #1-c. What is its result? Why it produced such output?

```
ggplot(mpg, aes(x = displ, y = hwy, color = year)) +
  geom_point() +
  labs(title = "Relationship between Engine Displacement and Highway MPG (Mapped with Year)",
       x = "Engine Displacement",
       y = "Highway MPG") +
  theme_minimal() +
  scale_color_gradient(low = "blue", high = "red")
```


Relationship between Engine Displacement and Highway MPG (Mapped with year)



6. Import the traffic.csv onto your R environment.

a. How many numbers of observation does it have? What are the variables of the traffic dataset the Show your answer.

```
library(readr)
traffic <- read_csv("traffic.csv")

## Rows: 48120 Columns: 4
## -- Column specification -----
## Delimiter: ","
## dbl (3): Junction, Vehicles, ID
## dtm (1): DateTime
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
varoftraffic <- colnames(traffic)

lengthVT <- length(varoftraffic)

# Number of Variables is 4
```

6b. subset the traffic dataset into junctions. What is the R codes and its output?

```
selected_junctions <- c(1, 2, 3, 4)

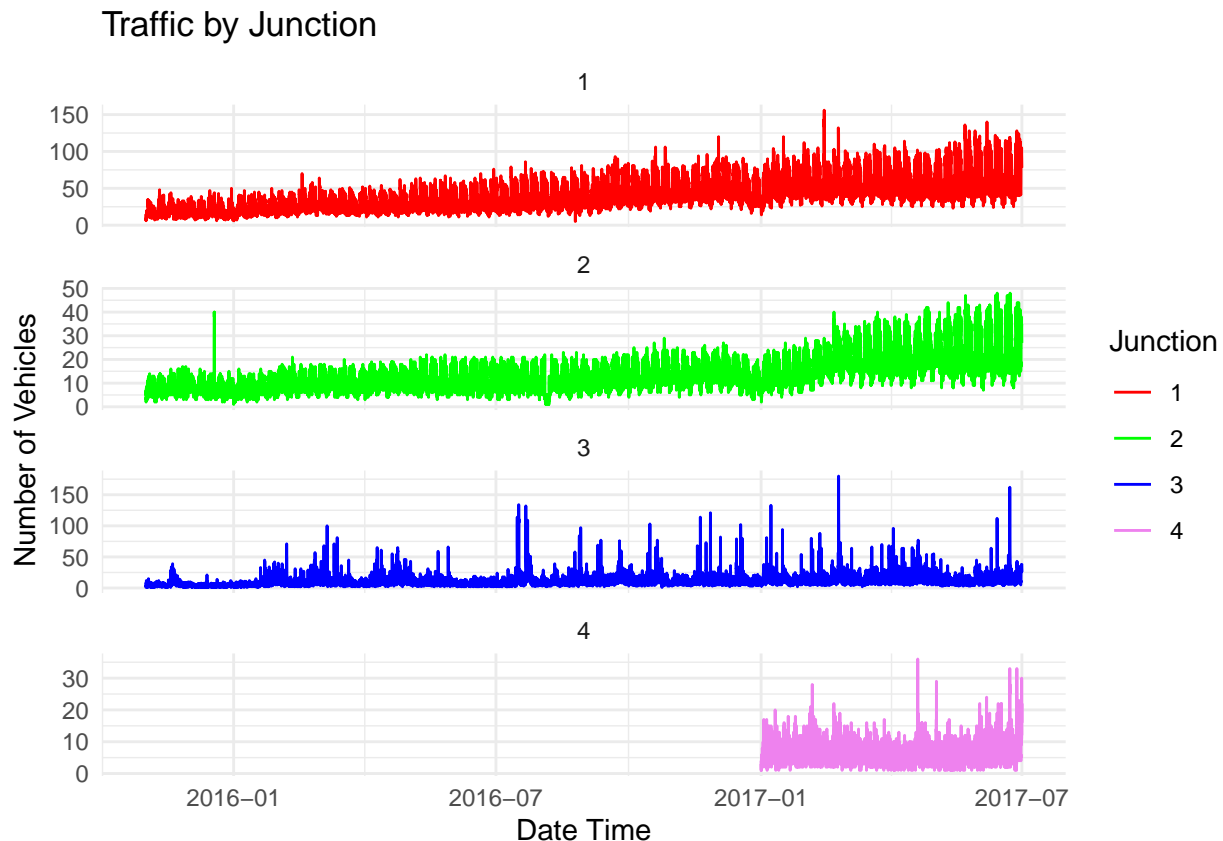
junction_data <- traffic[traffic$Junction %in% selected_junctions, ]
junction_data
```

```
## # A tibble: 48,120 x 4
##   DateTime      Junction Vehicles      ID
##   <dtm>         <dbl>     <dbl>   <dbl>
## 1 2015-11-01 00:00:00      1      15 20151101001
## 2 2015-11-01 01:00:00      1      13 20151101011
## 3 2015-11-01 02:00:00      1      10 20151101021
## 4 2015-11-01 03:00:00      1       7 20151101031
## 5 2015-11-01 04:00:00      1       9 20151101041
## 6 2015-11-01 05:00:00      1       6 20151101051
## 7 2015-11-01 06:00:00      1       9 20151101061
## 8 2015-11-01 07:00:00      1       8 20151101071
## 9 2015-11-01 08:00:00      1      11 20151101081
## 10 2015-11-01 09:00:00      1      12 20151101091
## # i 48,110 more rows
```

6c. Plot each junction in a using `geom_line()`. Show your solution and output.

```
library(ggplot2)
```

```
ggplot(traffic, aes(x = DateTime, y = Vehicles, color = as.factor(Junction))) +
  geom_line() +
  scale_color_manual(values = c("red", "green", "blue", "violet")) + # Adjust with your colors
  labs(title = "Traffic by Junction",
       x = "Date Time",
       y = "Number of Vehicles") +
  theme_minimal() +
  facet_wrap(~Junction, scales = "free_y", ncol = 1) +
  guides(color = guide_legend(title = "Junction"))
```



7. From alexa_file.xlsx, import it to your environment

```
library(readxl)
alexa_file <- read_excel("alexa_file.xlsx")
```

7a. How many observations does alexa_file has? What about the number of columns? Show your solution and answer.

```
obsofalex <- nrow(alexa_file)
obsofalex
```

```
## [1] 3150
```

```
colofalex <- ncol(alexa_file)
colofalex
```

```
## [1] 5
```

```
cat("Number of Observations:", obsofalex, "\n")
```

```
## Number of Observations: 3150
```

```
cat("Number of Columns:", colofalex, "\n")
```

```
## Number of Columns: 5
```

7b. group the variations and get the total of each variations. Use dplyr package. Show solution and answer.

```
library(dplyr)
```

```
groupofVar <- alexa_file %>%
  group_by(variation) %>%
  summarise(total_count = n())
groupofVar
```

```
## # A tibble: 16 x 2
##   variation                total_count
##   <chr>                  <int>
## 1 Black                    261
## 2 Black Dot                516
## 3 Black Plus               270
## 4 Black Show               265
## 5 Black Spot               241
## 6 Charcoal Fabric          430
## 7 Configuration: Fire TV Stick 350
## 8 Heather Gray Fabric      157
## 9 Oak Finish                14
## 10 Sandstone Fabric         90
## 11 Walnut Finish            9
## 12 White                    91
## 13 White Dot                184
## 14 White Plus               78
## 15 White Show               85
## 16 White Spot              109
```

7c. Plot the variations using the ggplot() function. What did you observe? Complete the details of the graph. Show solution and answer.

```
result <- alexa_file %>%
  group_by(variation) %>%
```

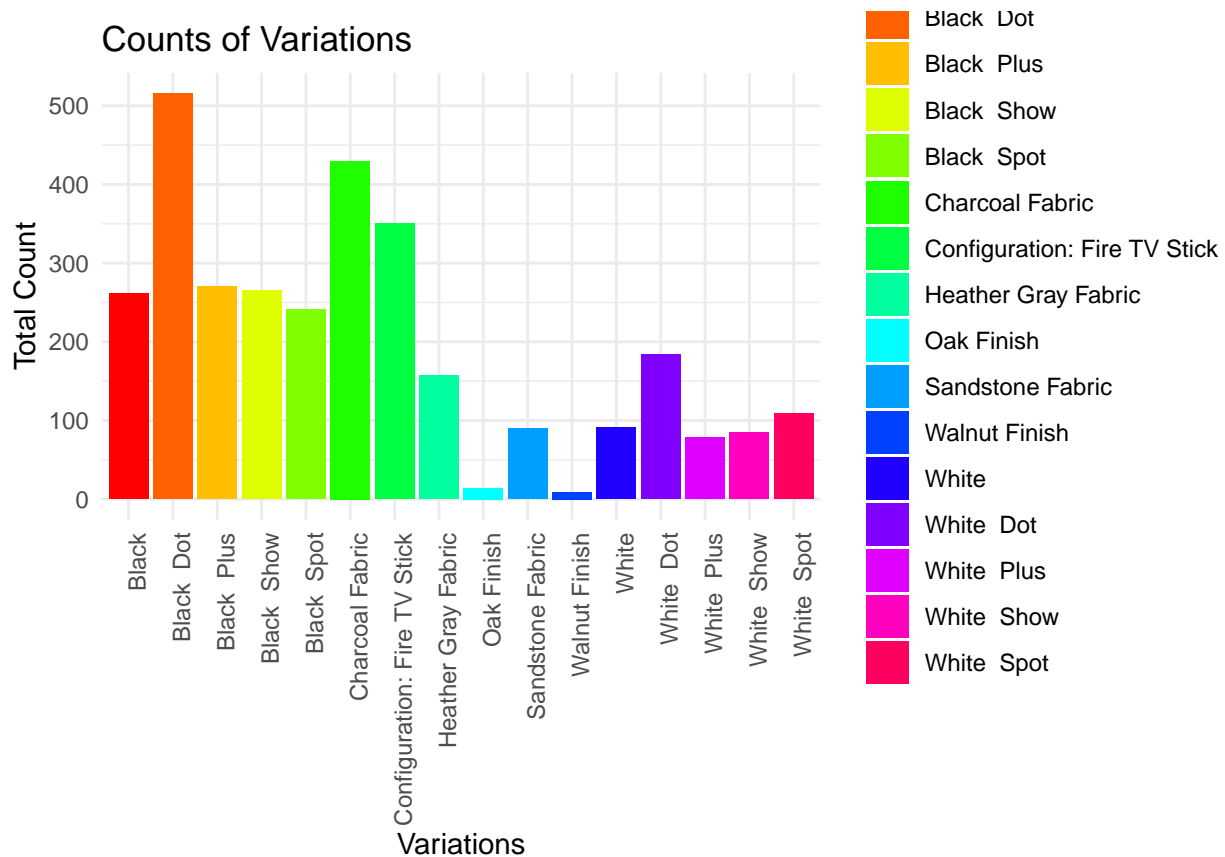
```

summarise(total_count = n())

plotofVar <- ggplot(result, aes(x = variation, y = total_count, fill = variation)) +
  geom_bar(stat = "identity") +
  scale_fill_manual(values = rainbow(n = nrow(result))) + # Rainbow colors
  labs(title = "Counts of Variations",
       x = "Variations",
       y = "Total Count") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 90, hjust = 1))

```

plotofVar



7d. Plot a `geom_line()` with the date and the number of verified reviews. Complete the details of the graphs. Show your answer and solution.

```

alexa_file$date <- as.Date(alexa_file$date)

alexa_file$month <- format(alexa_file$date, "%m")

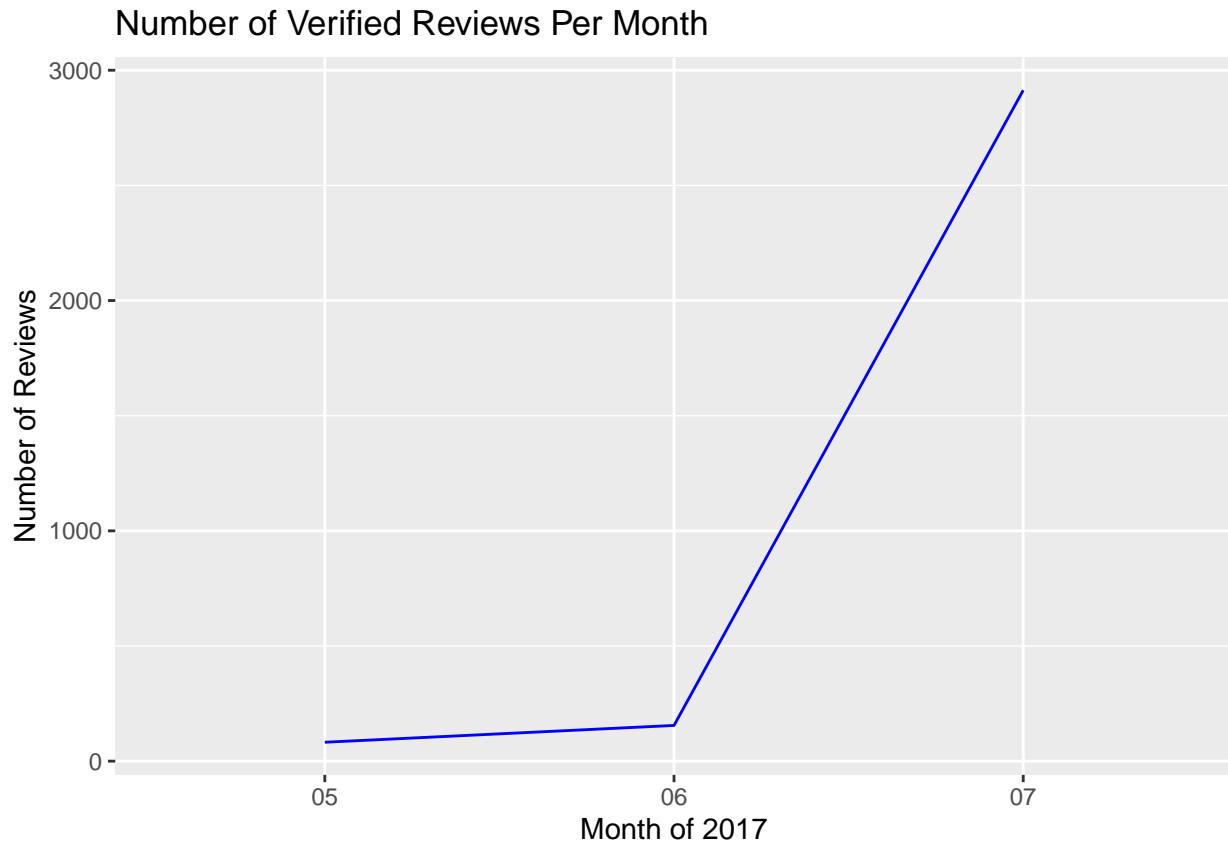
countsss <- alexa_file %>%
  group_by(month) %>%
  summarise(num_reviews = n())

monthly_review_counts <- table(countsss)

ggplot(countsss, aes(x = month, y = num_reviews, group = 1)) +

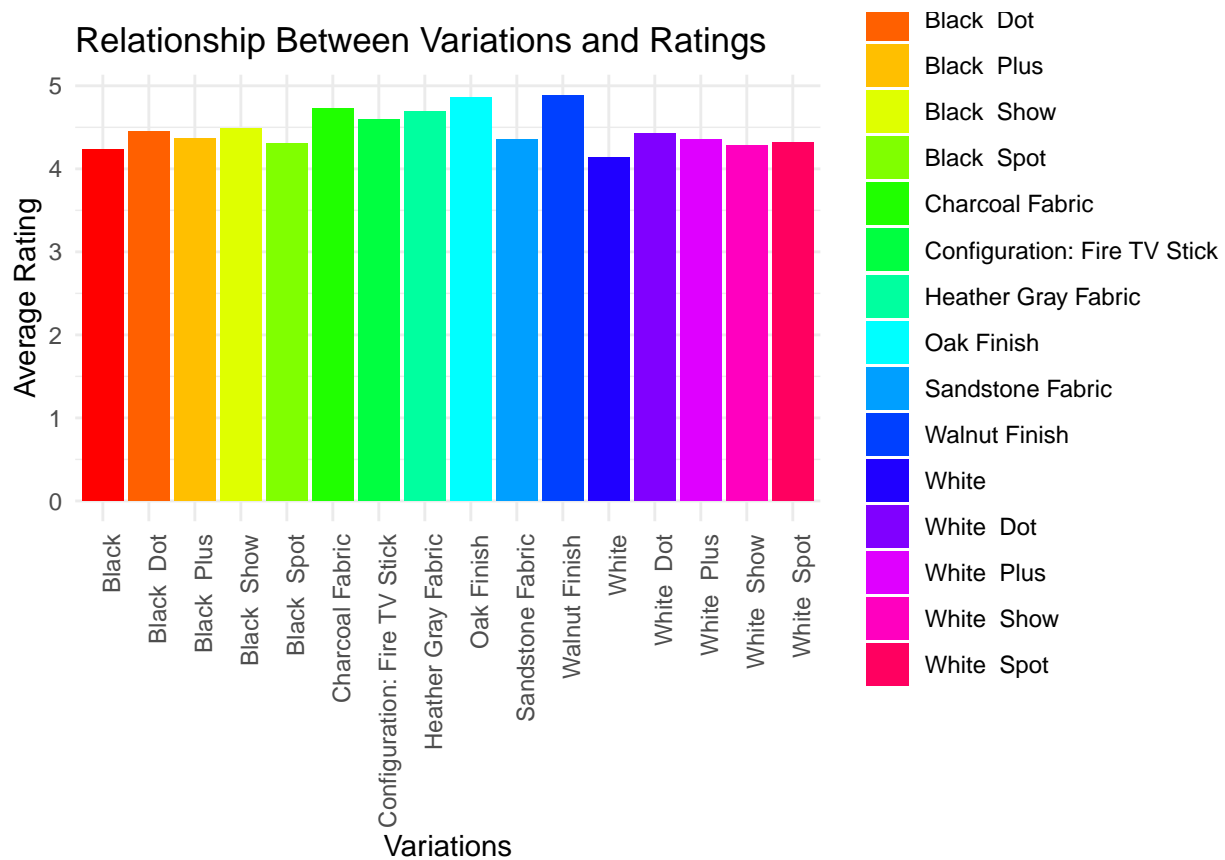
```

```
geom_line(color = "blue") +
labs(title = "Number of Verified Reviews Per Month",
      x = "Month of 2017", y = "Number of Reviews")
```



7e. Get the relationship of variations and ratings. Which variations got the most highest in rating? Plot a graph to show its relationship. Show your solution and answer.

```
ggplot(alexa_file, aes(x = variation, y = rating, fill = variation)) +
  geom_bar(stat = "summary", fun = "mean", position = "dodge") +
  scale_fill_manual(values = rainbow(n = length(unique(alexa_file$variation)))) + # Rainbow colors
  labs(title = "Relationship Between Variations and Ratings",
        x = "Variations",
        y = "Average Rating") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 90, hjust = 1))
```



```
variation_ratings <- alexa_file %>%
  group_by(variation) %>%
  summarise(avg_rating = mean(rating, na.rm = TRUE))

topvar <- variation_ratings %>%
  filter(avg_rating == max(avg_rating)) %>%
  pull(variation)

cat("Variation with the Highest Average Rating:", topvar, "\n")
```

```
## Variation with the Highest Average Rating: Walnut Finish
```