

# Worksheet 4c

Carl

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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))

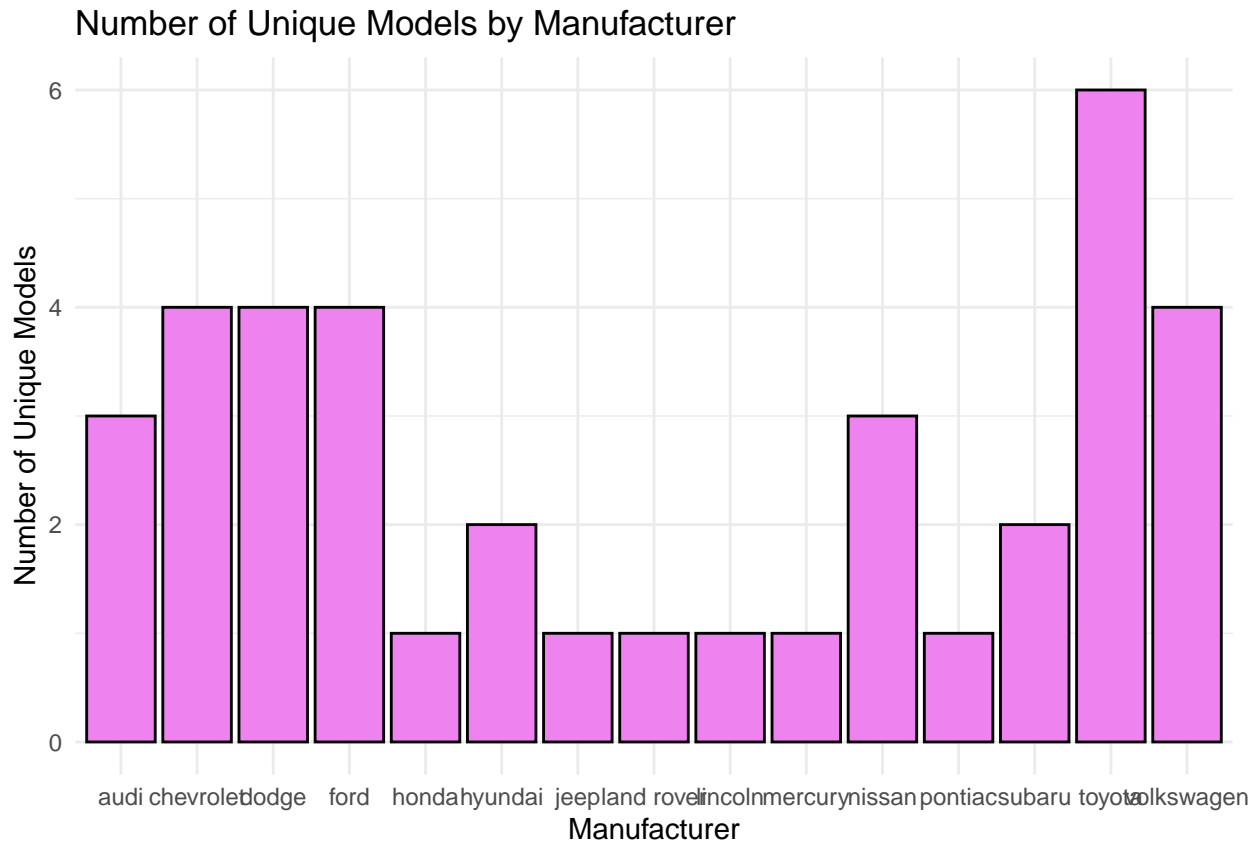
barplot(manufacturer_counts$unique_model_count, names.arg = manufacturer_counts$manufacturer,
        col = "violet", border = "black", main = "Number of Unique Models by Manufacturer",
        xlab = "Manufacturer", ylab = "Number of Unique Models",
        las = 2)
```

```
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.

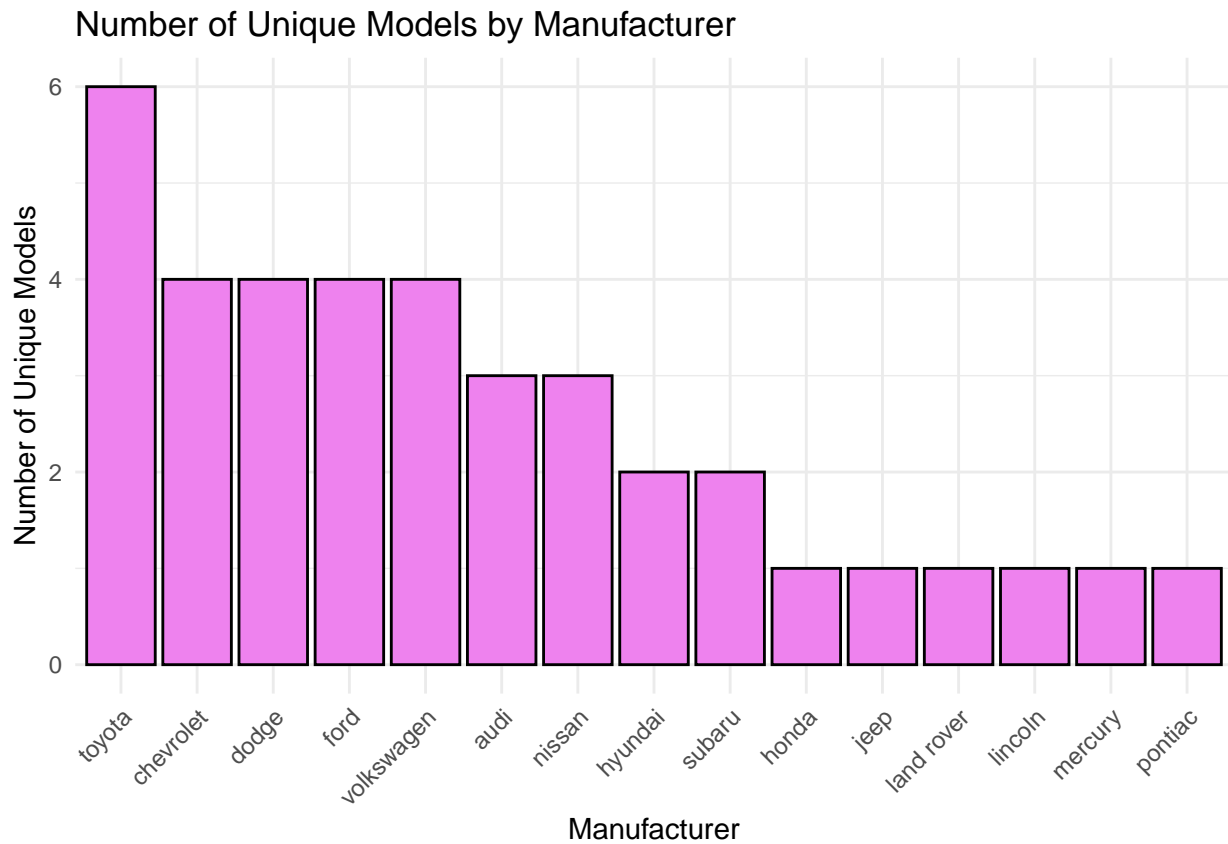
```
manufacturer_model_counts <- mpg %>%
  group_by(mpg$manufacturer, mpg$model) %>%
  summarise(count = n(), .groups = "drop")
```

```
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))
```

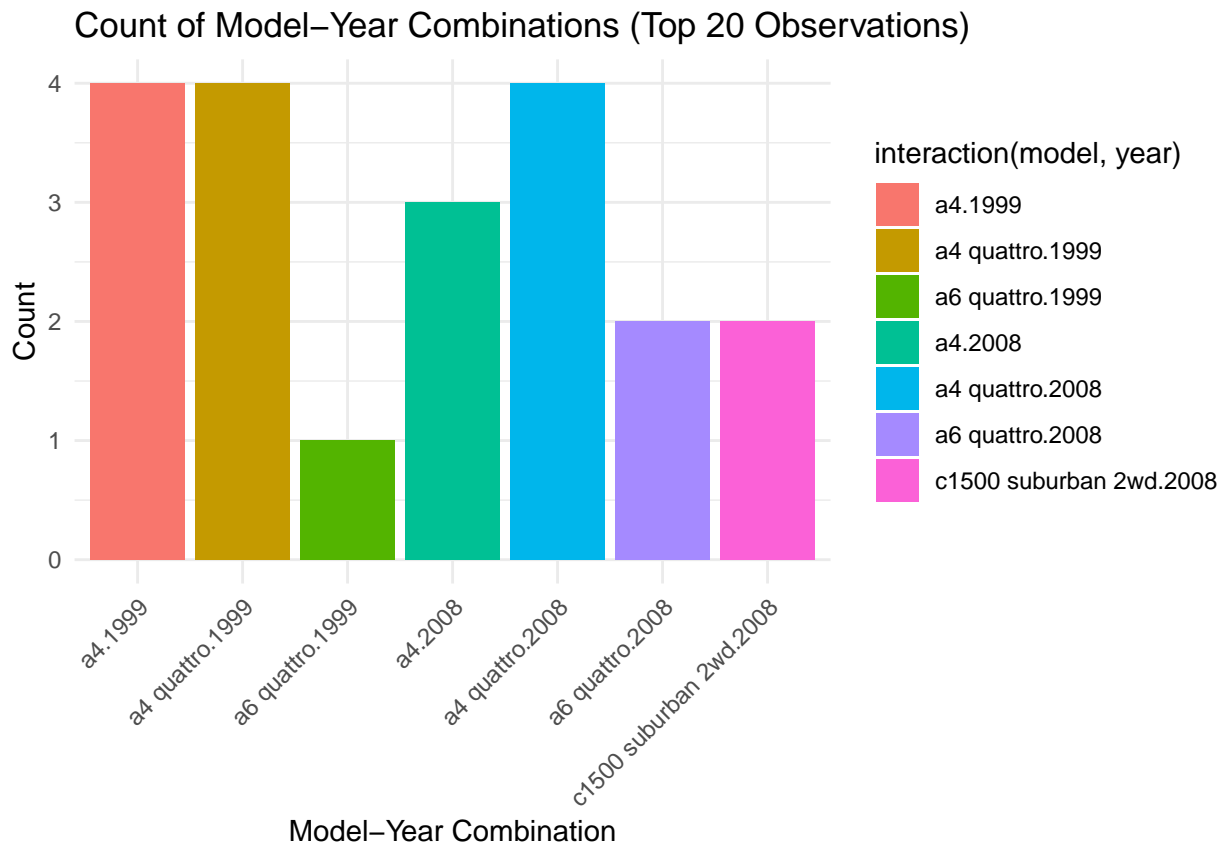


3a. What does `ggplot(mpg, aes(model, manufacturer)) + geom_point()` show?

```
ggplot(mpg, aes(model, manufacturer)) +
  geom_count()
```







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

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

```
cars_per_model
```

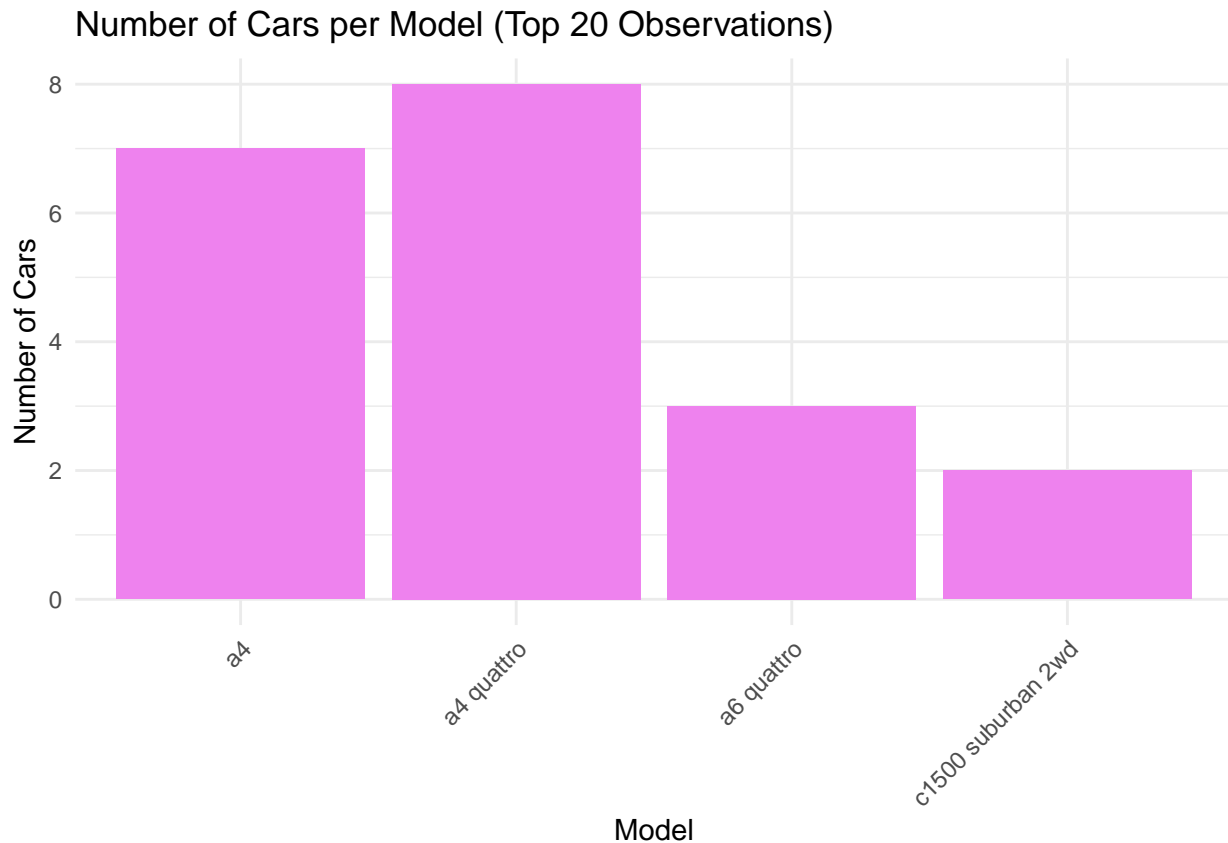
```
## # A tibble: 38 x 2
##   model                number_of_cars
##   <chr>                  <int>
## 1 4runner 4wd              6
## 2 a4                      7
## 3 a4 quattro              8
## 4 a6 quattro              3
## 5 altima                  6
## 6 c1500 suburban 2wd      5
## 7 camry                   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.

```
top_20 <- head(mpg, 20)
```

```
cars_per_model <- top_20 %>%
  group_by(model) %>%
  summarise(number_of_cars = n())

ggplot(cars_per_model, aes(x = model, y = number_of_cars)) +
  geom_bar(stat = "identity", fill = "violet") +
  labs(title = "Number of Cars per Model (Top 20 Observations)",
       x = "Model",
       y = "Number of Cars") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



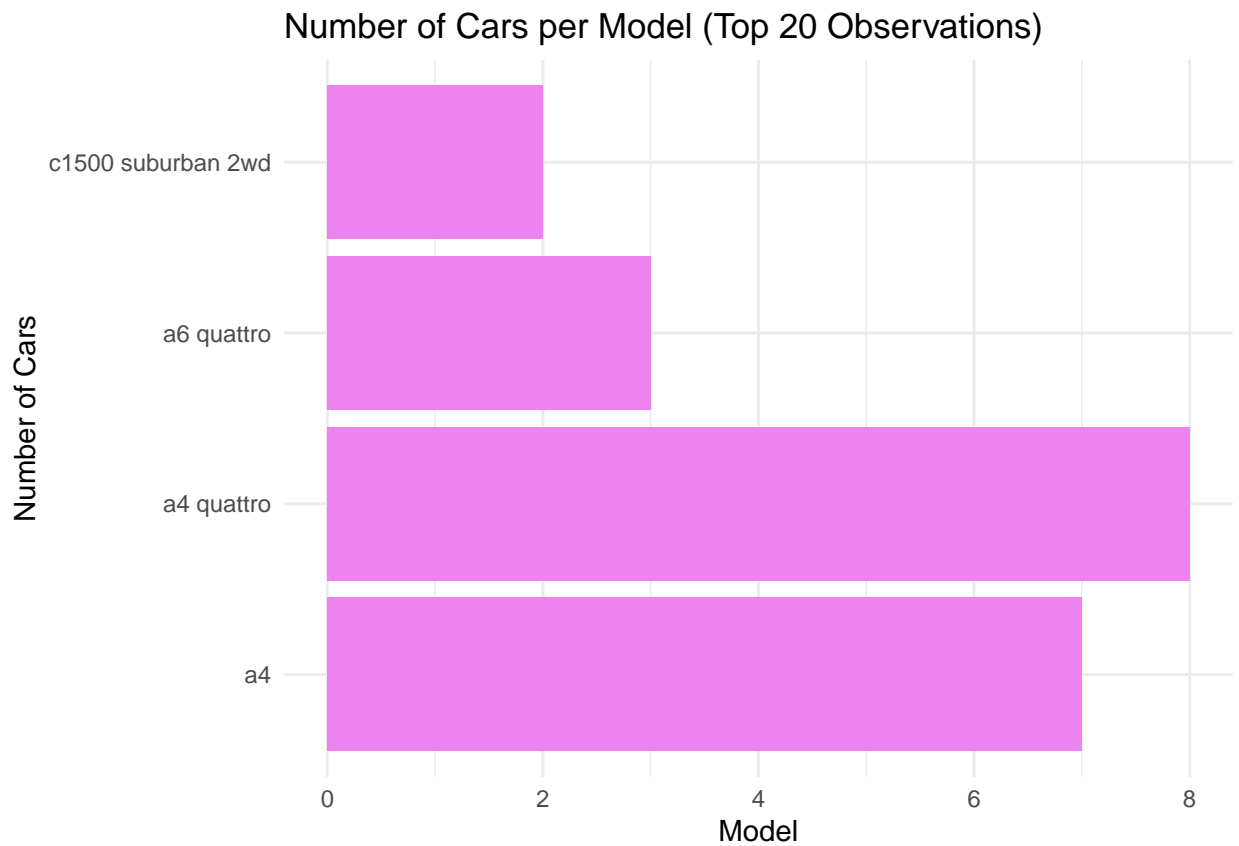
4b. Plot using the `geom_bar()` + `coord_flip()` just like what is shown below. Show codes and its result.

```
top_20 <- head(mpg, 20)

cars_per_model <- top_20 %>%
  group_by(model) %>%
  summarise(number_of_cars = n())

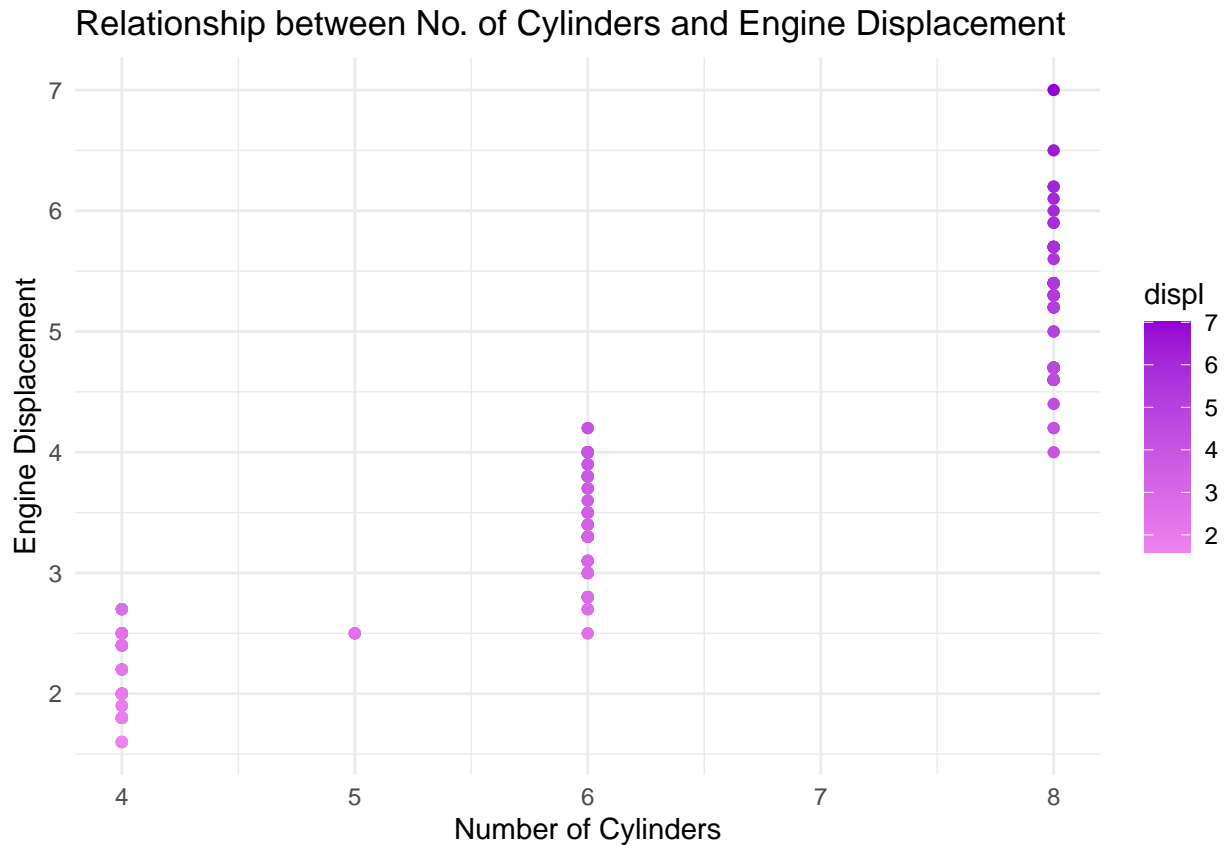
# Plot the result using geom_bar() with coord_flip()
ggplot(cars_per_model, aes(x = model, y = number_of_cars)) +
  geom_bar(stat = "identity", fill = "violet") +
  labs(title = "Number of Cars per Model (Top 20 Observations)",
       x = "Number of Cars",
       y = "Model") +
  theme_minimal() +
```

```
coord_flip()
```



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

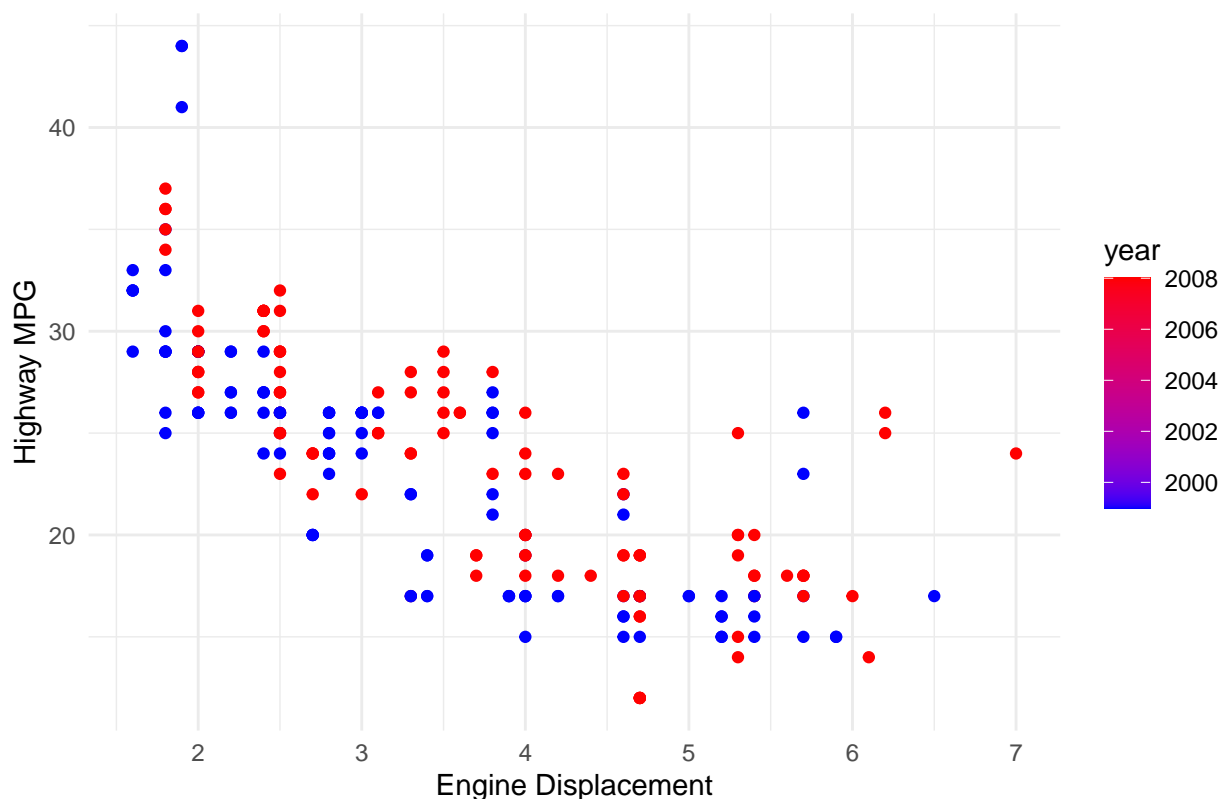
```
# number of cylenders increases the displacement also increases
```

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)

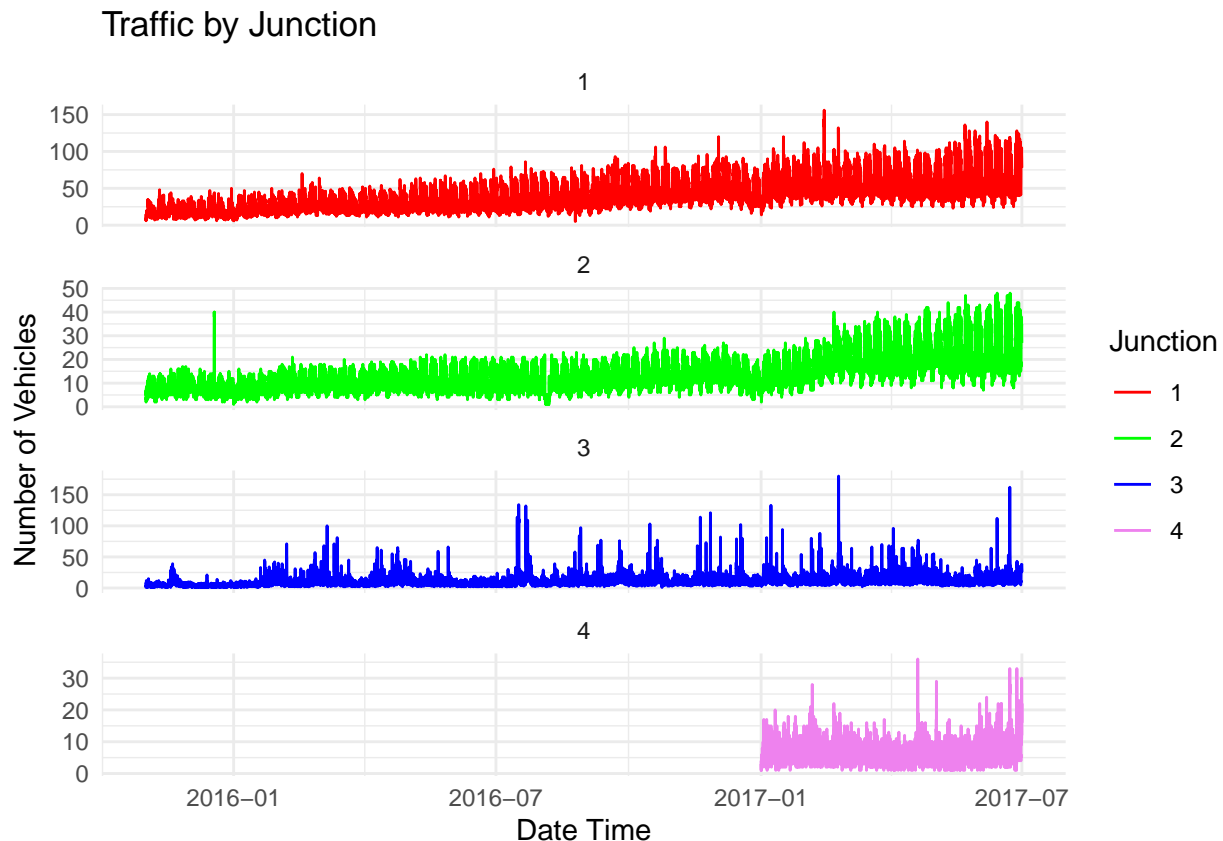
junction_data <- traffic[traffic$Junction %in% selected_junctions, ]
junction_data
```

```
## # 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) %>%
```

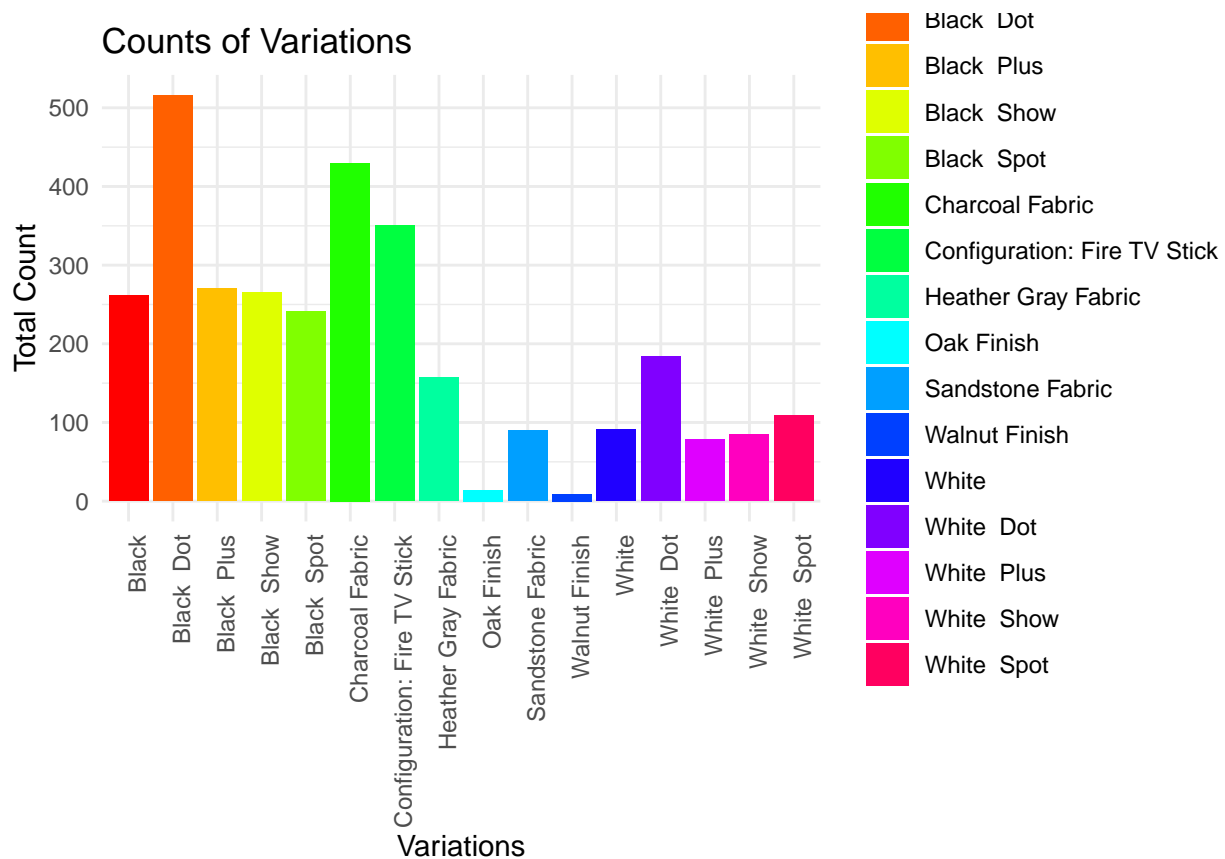
```

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")

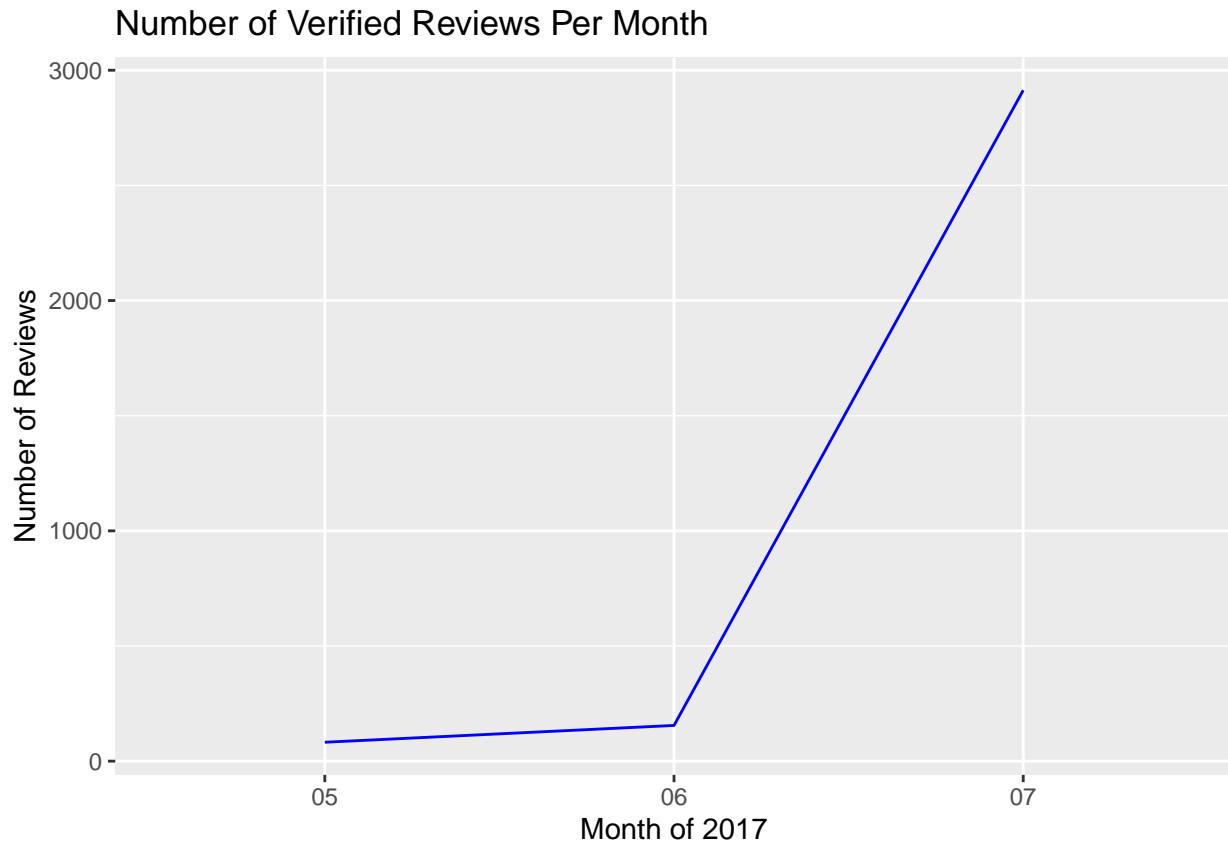
countsss <- alexa_file %>%
  group_by(month) %>%
  summarise(num_reviews = n())

monthly_review_counts <- table(countsss)

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

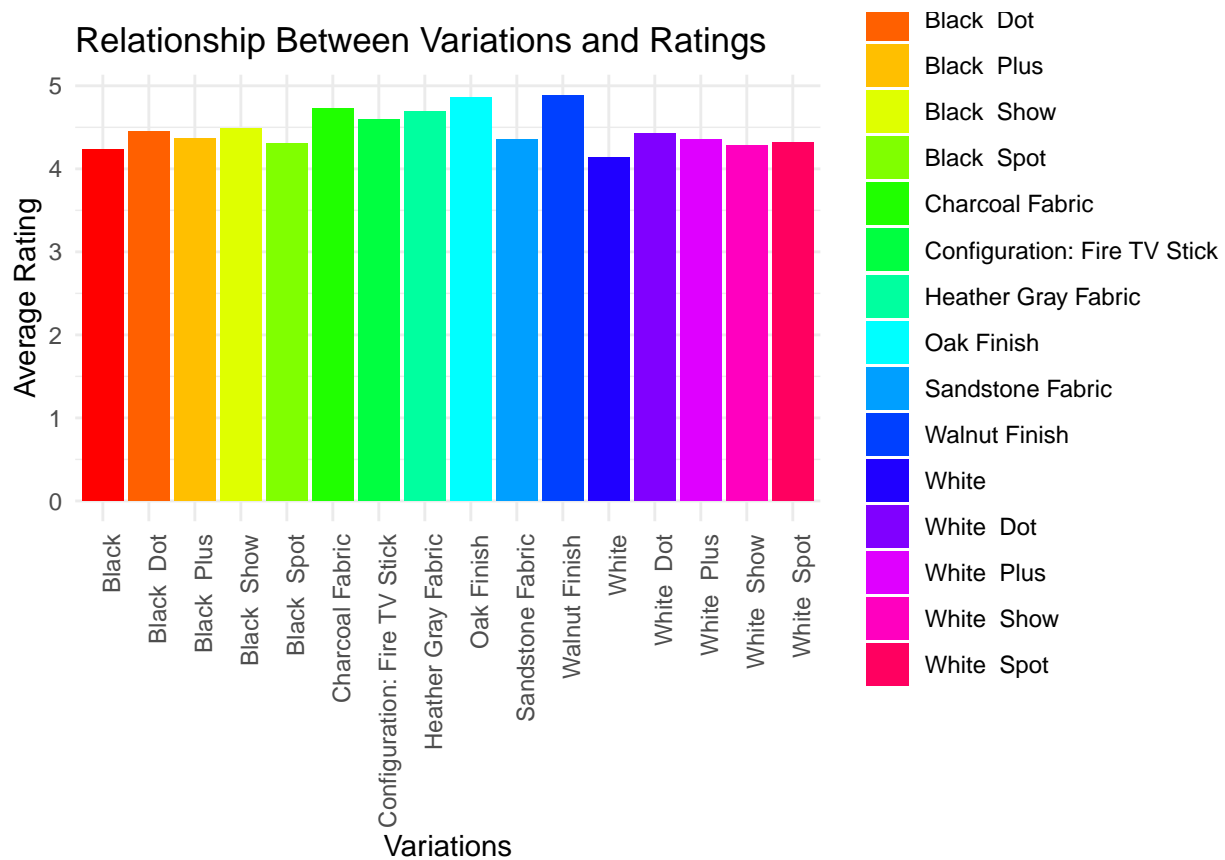
```

```
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
```