

Big Data Assignment 1

Due Date: Wednesday, Feb. 20, 2019, before class

A note from your TAs:

Hi! We recognize that this file is a large file and may be a bit overwhelming at first – don't worry! We'll be here to help you with any and all questions you may have. With that being said, there are a couple of house keeping notes:

1. For some of the questions below, you'll see that we've included code chunks underneath the question. This is where you'll type in the code that will be grade. **Please do not** modify the chunk's properties (aka the `results = FALSE`) that you'll see at the top of each chunk. Even with these modifications, you can still run your code and view your specific results.
2. You will also see `
` pieces throughout the document. **Please do not** delete these tags, as they are for formatting purposes. If you want to add text to your responses, please ensure that there is an empty line between your last line and any of the `
` tags.
3. When you submit your assignment, please just submit this file and rename it `bigdata_asst1_lastname.Rmd`

Thank you so much for reading this, and good luck with the assignment!

Question 1: Using R built-in datasets.

- a. Use the R help function to identify 2 built-in datasets. Provide a 1-2 sentence description of one of them. Write down the code to load a built-in dataset for R.

```
#####Q1a#####
```

```
#data() # Outputs all the inbuilt- datasets
```

```
AirPassengers
```

```
##      Jan Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec
## 1949 112 118 132 129 121 135 148 148 136 119 104 118
## 1950 115 126 141 135 125 149 170 170 158 133 114 140
## 1951 145 150 178 163 172 178 199 199 184 162 146 166
## 1952 171 180 193 181 183 218 230 242 209 191 172 194
## 1953 196 196 236 235 229 243 264 272 237 211 180 201
## 1954 204 188 235 227 234 264 302 293 259 229 203 229
```

```

## 1955 242 233 267 269 270 315 364 347 312 274 237 278
## 1956 284 277 317 313 318 374 413 405 355 306 271 306
## 1957 315 301 356 348 355 422 465 467 404 347 305 336
## 1958 340 318 362 348 363 435 491 505 404 359 310 337
## 1959 360 342 406 396 420 472 548 559 463 407 362 405
## 1960 417 391 419 461 472 535 622 606 508 461 390 432

```

Monthly Airline Passenger Numbers 1949-1960

USJudgeRatings

##	CONT	INTG	DMNR	DILG	CFMG	DECI	PREP	FAMI	ORAL	WRIT	PHYS
## AARONSON, L.H.	5.7	7.9	7.7	7.3	7.1	7.4	7.1	7.1	7.1	7.0	8.3
## ALEXANDER, J.M.	6.8	8.9	8.8	8.5	7.8	8.1	8.0	8.0	7.8	7.9	8.5
## ARMENTANO, A.J.	7.2	8.1	7.8	7.8	7.5	7.6	7.5	7.5	7.3	7.4	7.9
## BERDON, R.I.	6.8	8.8	8.5	8.8	8.3	8.5	8.7	8.7	8.4	8.5	8.8
## BRACKEN, J.J.	7.3	6.4	4.3	6.5	6.0	6.2	5.7	5.7	5.1	5.3	5.5
## BURNS, E.B.	6.2	8.8	8.7	8.5	7.9	8.0	8.1	8.0	8.0	8.0	8.6
## CALLAHAN, R.J.	10.6	9.0	8.9	8.7	8.5	8.5	8.5	8.5	8.6	8.4	9.1
## COHEN, S.S.	7.0	5.9	4.9	5.1	5.4	5.9	4.8	5.1	4.7	4.9	6.8
## DALY, J.J.	7.3	8.9	8.9	8.7	8.6	8.5	8.4	8.4	8.4	8.5	8.8
## DANNEHY, J.F.	8.2	7.9	6.7	8.1	7.9	8.0	7.9	8.1	7.7	7.8	8.5
## DEAN, H.H.	7.0	8.0	7.6	7.4	7.3	7.5	7.1	7.2	7.1	7.2	8.4
## DEVITA, H.J.	6.5	8.0	7.6	7.2	7.0	7.1	6.9	7.0	7.0	7.1	6.9
## DRISCOLL, P.J.	6.7	8.6	8.2	6.8	6.9	6.6	7.1	7.3	7.2	7.2	8.1
## GRILLO, A.E.	7.0	7.5	6.4	6.8	6.5	7.0	6.6	6.8	6.3	6.6	6.2
## HADDEN, W.L. JR.	6.5	8.1	8.0	8.0	7.9	8.0	7.9	7.8	7.8	7.8	8.4
## HAMILL, E.C.	7.3	8.0	7.4	7.7	7.3	7.3	7.3	7.2	7.1	7.2	8.0
## HEALEY, A.H.	8.0	7.6	6.6	7.2	6.5	6.5	6.8	6.7	6.4	6.5	6.9
## HULL, T.C.	7.7	7.7	6.7	7.5	7.4	7.5	7.1	7.3	7.1	7.3	8.1
## LEVINE, I.	8.3	8.2	7.4	7.8	7.7	7.7	7.7	7.8	7.5	7.6	8.0
## LEVISTER, R.L.	9.6	6.9	5.7	6.6	6.9	6.6	6.2	6.0	5.8	5.8	7.2
## MARTIN, L.F.	7.1	8.2	7.7	7.1	6.6	6.6	6.7	6.7	6.8	6.8	7.5
## MCGRATH, J.F.	7.6	7.3	6.9	6.8	6.7	6.8	6.4	6.3	6.3	6.3	7.4
## MIGNONE, A.F.	6.6	7.4	6.2	6.2	5.4	5.7	5.8	5.9	5.2	5.8	4.7
## MISSAL, H.M.	6.2	8.3	8.1	7.7	7.4	7.3	7.3	7.3	7.2	7.3	7.8
## MULVEY, H.M.	7.5	8.7	8.5	8.6	8.5	8.4	8.5	8.5	8.4	8.4	8.7
## NARUK, H.J.	7.8	8.9	8.7	8.9	8.7	8.8	8.9	9.0	8.8	8.9	9.0
## O'BRIEN, F.J.	7.1	8.5	8.3	8.0	7.9	7.9	7.8	7.8	7.8	7.7	8.3
## O'SULLIVAN, T.J.	7.5	9.0	8.9	8.7	8.4	8.5	8.4	8.3	8.3	8.3	8.8
## PASKEY, L.	7.5	8.1	7.7	8.2	8.0	8.1	8.2	8.4	8.0	8.1	8.4
## RUBINOW, J.E.	7.1	9.2	9.0	9.0	8.4	8.6	9.1	9.1	8.9	9.0	8.9
## SADEN, G.A.	6.6	7.4	6.9	8.4	8.0	7.9	8.2	8.4	7.7	7.9	8.4
## SATANIELLO, A.G.	8.4	8.0	7.9	7.9	7.8	7.8	7.6	7.4	7.4	7.4	8.1
## SHEA, D.M.	6.9	8.5	7.8	8.5	8.1	8.2	8.4	8.5	8.1	8.3	8.7
## SHEA, J.F. JR.	7.3	8.9	8.8	8.7	8.4	8.5	8.5	8.5	8.4	8.4	8.8
## SIDOR, W.J.	7.7	6.2	5.1	5.6	5.6	5.9	5.6	5.6	5.3	5.5	6.3
## SPEZIALE, J.A.	8.5	8.3	8.1	8.3	8.4	8.2	8.2	8.1	7.9	8.0	8.0
## SPONZO, M.J.	6.9	8.3	8.0	8.1	7.9	7.9	7.9	7.7	7.6	7.7	8.1
## STAPLETON, J.F.	6.5	8.2	7.7	7.8	7.6	7.7	7.7	7.7	7.5	7.6	8.5

[illegible]

Lawyers' Ratings of State Judges in the US Superior Court

Use the R dataset `Seatbelts` to answer the following:

- b. What does the dataset contain? There are several ways to figure this out. Use two different ways. One way is just to type `Seatbelts` at the R prompt. Other commands to explore are `str()`, `summary()`, `dim()`, `nrow()`, and `ncol()` where you put the name of the database within the parenthesis. Apply each of these functions to `Seatbelts`. Furthermore, apply `is.na()` and `is.null()` to check to see if there are any missing data from our datasets.

#####Q1b#####

Seatbelts

##		DriversKilled	drivers	front	rear	kms	PetrolPrice	VanKilled	law
##	Jan 1969	107	1687	867	269	9059	0.10297181	12	0
##	Feb 1969	97	1508	825	265	7685	0.10236300	6	0
##	Mar 1969	102	1507	806	319	9963	0.10206249	12	0
##	Apr 1969	87	1385	814	407	10955	0.10087330	8	0
##	May 1969	119	1632	991	454	11823	0.10101967	10	0
##	Jun 1969	106	1511	945	427	12391	0.10058119	13	0
##	Jul 1969	110	1559	1004	522	13460	0.10377398	11	0
##	Aug 1969	106	1630	1091	536	14055	0.10407640	6	0
##	Sep 1969	107	1579	958	405	12106	0.10377398	10	0
##	Oct 1969	134	1653	850	437	11372	0.10302640	16	0
##	Nov 1969	147	2152	1109	434	9834	0.10273011	13	0
##	Dec 1969	180	2148	1113	437	9267	0.10199719	14	0
##	Jan 1970	125	1752	925	316	9130	0.10127456	14	0
##	Feb 1970	134	1765	903	311	8933	0.10070398	6	0
##	Mar 1970	110	1717	1006	351	11000	0.10013961	8	0
##	Apr 1970	102	1558	892	362	10733	0.09862110	11	0
##	May 1970	103	1575	990	486	12912	0.09834929	7	0
##	Jun 1970	111	1520	866	429	12926	0.09808018	13	0
##	Jul 1970	120	1805	1095	551	13990	0.09727921	13	0
##	Aug 1970	129	1800	1204	646	14926	0.09741062	11	0
##	Sep 1970	122	1719	1029	456	12900	0.09742524	11	0
##	Oct 1970	183	2008	1147	475	12034	0.09638063	14	0
##	Nov 1970	169	2242	1171	456	10643	0.09573896	16	0
##	Dec 1970	190	2478	1299	468	10742	0.09510631	14	0
##	Jan 1971	134	2030	944	356	10266	0.09673597	17	0
##	Feb 1971	108	1655	874	271	10281	0.09610922	16	0
##	Mar 1971	104	1693	840	354	11527	0.09536725	15	0
##	Apr 1971	117	1623	893	427	12281	0.09470959	13	0
##	May 1971	157	1805	1007	465	13587	0.09411762	13	0
##	Jun 1971	148	1746	973	440	13049	0.09353215	15	0
##	Jul 1971	130	1795	1097	539	16055	0.09295405	12	0
##	Aug 1971	140	1926	1194	646	15220	0.09283979	6	0

## Sep 1971	136	1619	988	457	13824	0.09272474	9	0
## Oct 1971	140	1992	1077	446	12729	0.09226965	13	0
## Nov 1971	187	2233	1045	402	11467	0.09170669	14	0
## Dec 1971	150	2192	1115	441	11351	0.09126207	15	0
## Jan 1972	159	2080	1005	359	10803	0.09071160	14	0
## Feb 1972	143	1768	857	334	10548	0.09027633	3	0
## Mar 1972	114	1835	879	312	12368	0.08995192	12	0
## Apr 1972	127	1569	887	427	13311	0.08909964	13	0
## May 1972	159	1976	1075	434	13885	0.08867919	12	0
## Jun 1972	156	1853	1121	486	14088	0.08815929	8	0
## Jul 1972	138	1965	1190	569	16932	0.08890206	8	0
## Aug 1972	120	1689	1058	523	16164	0.08818133	15	0
## Sep 1972	117	1778	939	418	14883	0.08894029	8	0
## Oct 1972	170	1976	1074	452	13532	0.08772661	5	0
## Nov 1972	168	2397	1089	462	12220	0.08742885	17	0
## Dec 1972	198	2654	1208	497	12025	0.08703543	14	0
## Jan 1973	144	2097	903	354	11692	0.08644992	13	0
## Feb 1973	146	1963	916	347	11081	0.08587264	5	0
## Mar 1973	109	1677	787	276	13745	0.08539822	8	0
## Apr 1973	131	1941	1114	472	14382	0.08382198	5	0
## May 1973	151	2003	1014	487	14391	0.08459078	12	0
## Jun 1973	140	1813	1022	505	15597	0.08413690	11	0
## Jul 1973	153	2012	1114	619	16834	0.08377841	13	0
## Aug 1973	140	1912	1132	640	17282	0.08351074	15	0
## Sep 1973	161	2084	1111	559	15779	0.08280639	11	0
## Oct 1973	168	2080	1008	453	13946	0.08117889	11	0
## Nov 1973	152	2118	916	418	12701	0.08285361	10	0
## Dec 1973	136	2150	992	419	10431	0.09419012	13	0
## Jan 1974	113	1608	731	262	11616	0.09239984	8	0
## Feb 1974	100	1503	665	299	10808	0.10816148	6	0
## Mar 1974	103	1548	724	303	12421	0.10721169	8	0
## Apr 1974	103	1382	744	401	13605	0.11404297	14	0
## May 1974	121	1731	910	413	14455	0.11245412	12	0
## Jun 1974	134	1798	883	426	15019	0.11131625	14	0
## Jul 1974	133	1779	900	516	15662	0.11030125	13	0
## Aug 1974	129	1887	1057	600	16745	0.10819718	9	0
## Sep 1974	144	2004	1076	459	14717	0.10702744	4	0
## Oct 1974	154	2077	919	443	13756	0.10494698	13	0
## Nov 1974	156	2092	920	412	12531	0.11935775	6	0
## Dec 1974	163	2051	953	400	12568	0.11762190	15	0
## Jan 1975	122	1577	664	278	11249	0.13302742	12	0
## Feb 1975	92	1356	607	302	11096	0.13084524	16	0
## Mar 1975	117	1652	777	381	12637	0.12831848	7	0
## Apr 1975	95	1382	633	279	13018	0.12354745	12	0
## May 1975	96	1519	791	442	15005	0.11858681	10	0
## Jun 1975	108	1421	790	409	15235	0.11633748	9	0
## Jul 1975	108	1442	803	416	15552	0.11516148	9	0
## Aug 1975	106	1543	884	511	16905	0.11450120	6	0
## Sep 1975	140	1656	769	393	14776	0.11352298	7	0
## Oct 1975	114	1561	732	345	14104	0.11193018	13	0

## Nov 1975	158	1905	859	391	12854	0.11061053	14	0
## Dec 1975	161	2199	994	470	12956	0.11527439	13	0
## Jan 1976	102	1473	704	266	12177	0.11379349	14	0
## Feb 1976	127	1655	684	312	11918	0.11234958	11	0
## Mar 1976	125	1407	671	300	13517	0.11175347	11	0
## Apr 1976	101	1395	643	373	14417	0.10964252	10	0
## May 1976	97	1530	771	412	15911	0.10844090	4	0
## Jun 1976	112	1309	644	322	15589	0.10788494	8	0
## Jul 1976	112	1526	828	458	16543	0.10908477	9	0
## Aug 1976	113	1327	748	427	17925	0.10757145	10	0
## Sep 1976	108	1627	767	346	15406	0.10616402	10	0
## Oct 1976	128	1748	825	421	14601	0.10630000	5	0
## Nov 1976	154	1958	810	344	13107	0.10482531	13	0
## Dec 1976	162	2274	986	370	12268	0.10345175	12	0
## Jan 1977	112	1648	714	291	11972	0.10144992	10	0
## Feb 1977	79	1401	567	224	12028	0.10040232	9	0
## Mar 1977	82	1411	616	266	14033	0.09886203	7	0
## Apr 1977	127	1403	678	338	14244	0.10249615	5	0
## May 1977	108	1394	742	298	15287	0.10302743	10	0
## Jun 1977	110	1520	840	386	16954	0.10217891	5	0
## Jul 1977	123	1528	888	479	17361	0.09983664	6	0
## Aug 1977	103	1643	852	473	17694	0.09263669	8	0
## Sep 1977	97	1515	774	332	16222	0.09181496	6	0
## Oct 1977	140	1685	831	391	14969	0.09072430	12	0
## Nov 1977	165	2000	889	370	13624	0.09002121	15	0
## Dec 1977	183	2215	1046	431	13842	0.08933071	7	0
## Jan 1978	148	1956	889	366	12387	0.08844273	14	0
## Feb 1978	111	1462	626	250	11608	0.08835257	4	0
## Mar 1978	116	1563	808	355	15021	0.08675736	10	0
## Apr 1978	115	1459	746	304	14834	0.08499524	8	0
## May 1978	100	1446	754	379	16565	0.08456794	7	0
## Jun 1978	106	1622	865	440	16882	0.08443190	11	0
## Jul 1978	134	1657	980	500	18012	0.08435088	3	0
## Aug 1978	125	1638	959	511	18855	0.08360098	5	0
## Sep 1978	117	1643	856	384	17243	0.08341726	11	0
## Oct 1978	122	1683	798	366	16045	0.08274514	10	0
## Nov 1978	153	2050	942	432	14745	0.08523527	10	0
## Dec 1978	178	2262	1010	390	13726	0.08477030	7	0
## Jan 1979	114	1813	796	306	11196	0.08445892	10	0
## Feb 1979	94	1445	643	232	12105	0.08535212	11	0
## Mar 1979	128	1762	794	342	14723	0.08755921	9	0
## Apr 1979	119	1461	750	329	15582	0.09038292	7	0
## May 1979	111	1556	809	394	16863	0.09078329	8	0
## Jun 1979	110	1431	716	355	16758	0.10874278	13	0
## Jul 1979	114	1427	851	385	17434	0.11414223	8	0
## Aug 1979	118	1554	931	463	18359	0.11299293	5	0
## Sep 1979	115	1645	834	453	17189	0.11132071	8	0
## Oct 1979	132	1653	762	373	16909	0.10912623	7	0
## Nov 1979	153	2016	880	401	15380	0.10769846	12	0
## Dec 1979	171	2207	1077	466	15161	0.10760157	10	0

## Jan 1980	115	1665	748	306	14027	0.10377502	7	0
## Feb 1980	95	1361	593	263	14478	0.10711417	4	0
## Mar 1980	92	1506	720	323	16155	0.10737477	10	0
## Apr 1980	100	1360	646	310	16585	0.11169537	4	0
## May 1980	95	1453	765	424	18117	0.11063818	8	0
## Jun 1980	114	1522	820	403	17552	0.11185521	8	0
## Jul 1980	102	1460	807	406	18299	0.10974234	7	0
## Aug 1980	104	1552	885	466	19361	0.10819393	10	0
## Sep 1980	132	1548	803	381	17924	0.10625536	8	0
## Oct 1980	136	1827	860	369	17872	0.10419303	14	0
## Nov 1980	117	1737	825	378	16058	0.10193397	8	0
## Dec 1980	137	1941	911	392	15746	0.10279382	9	0
## Jan 1981	111	1474	704	284	15226	0.10476034	8	0
## Feb 1981	106	1458	691	316	14932	0.10400254	6	0
## Mar 1981	98	1542	688	321	16846	0.11665552	7	0
## Apr 1981	84	1404	714	358	16854	0.11516148	6	0
## May 1981	94	1522	814	378	18146	0.11298954	5	0
## Jun 1981	105	1385	736	382	17559	0.11386064	4	0
## Jul 1981	123	1641	876	433	18655	0.11911808	5	0
## Aug 1981	109	1510	829	506	19453	0.12448999	10	0
## Sep 1981	130	1681	818	428	17923	0.12322295	7	0
## Oct 1981	153	1938	942	479	17915	0.12067793	10	0
## Nov 1981	134	1868	782	370	16496	0.12104898	12	0
## Dec 1981	99	1726	823	349	13544	0.11696857	7	0
## Jan 1982	115	1456	595	238	13601	0.11275026	4	0
## Feb 1982	104	1445	673	285	15667	0.10807931	5	0
## Mar 1982	131	1456	660	324	17358	0.10883852	6	0
## Apr 1982	108	1365	676	346	18112	0.11129177	4	0
## May 1982	103	1487	755	410	18581	0.11130401	4	0
## Jun 1982	115	1558	815	411	18759	0.11545436	8	0
## Jul 1982	122	1488	867	496	20668	0.11476830	8	0
## Aug 1982	122	1684	933	534	21040	0.11720743	3	0
## Sep 1982	125	1594	798	396	18993	0.11907640	7	0
## Oct 1982	137	1850	950	470	18668	0.11796586	12	0
## Nov 1982	138	1998	825	385	16768	0.11744913	2	0
## Dec 1982	152	2079	911	411	16551	0.11698846	7	0
## Jan 1983	120	1494	619	281	16231	0.11261054	8	0
## Feb 1983	95	1057	426	300	15511	0.11365702	3	1
## Mar 1983	100	1218	475	318	18308	0.11314445	2	1
## Apr 1983	89	1168	556	391	17793	0.11849553	6	1
## May 1983	82	1236	559	398	19205	0.11796940	3	1
## Jun 1983	89	1076	483	337	19162	0.11768661	7	1
## Jul 1983	60	1174	587	477	20997	0.12005924	6	1
## Aug 1983	84	1139	615	422	20705	0.11943775	8	1
## Sep 1983	113	1427	618	495	18759	0.11888127	8	1
## Oct 1983	126	1487	662	471	19240	0.11846236	4	1
## Nov 1983	122	1483	519	368	17504	0.11801660	3	1
## Dec 1983	118	1513	585	345	16591	0.11770662	5	1
## Jan 1984	92	1357	483	296	16224	0.11777609	5	1
## Feb 1984	86	1165	434	319	16670	0.11479699	3	1

```
## Mar 1984      81    1282    513    349 18539  0.11573525      4    1
## Apr 1984      84    1110    548    375 19759  0.11535626      3    1
## May 1984      87    1297    586    441 19584  0.11481536      6    1
## Jun 1984      90    1185    522    465 19976  0.11477748      6    1
## Jul 1984      79    1222    601    472 21486  0.11493598      7    1
## Aug 1984      96    1284    644    521 21626  0.11479699      5    1
## Sep 1984     122    1444    643    429 20195  0.11409316      7    1
## Oct 1984     120    1575    641    408 19928  0.11646552      7    1
## Nov 1984     137    1737    711    490 18564  0.11602611      4    1
## Dec 1984     154    1763    721    491 18149  0.11606673      7    1
```

```
print('Variables of Seatbelts ')
```

```
## [1] "Variables of Seatbelts "
```

```
str(Seatbelts)
```

```
## Time-Series [1:192, 1:8] from 1969 to 1985: 107 97 102 87 119 106 110 106
107 134 ...
## - attr(*, "dimnames")=List of 2
## ..$ : NULL
## ..$ : chr [1:8] "DriversKilled" "drivers" "front" "rear" ...
```

```
print('Summary of Seatbelts ')
```

```
## [1] "Summary of Seatbelts "
```

```
summary(Seatbelts)
```

```
## DriversKilled      drivers      front      rear
## Min.   : 60.0    Min.   :1057    Min.   : 426.0    Min.   :224.0
## 1st Qu.:104.8    1st Qu.:1462    1st Qu.: 715.5    1st Qu.:344.8
## Median :118.5    Median :1631    Median : 828.5    Median :401.5
## Mean   :122.8    Mean   :1670    Mean   : 837.2    Mean   :401.2
## 3rd Qu.:138.0    3rd Qu.:1851    3rd Qu.: 950.8    3rd Qu.:456.2
## Max.   :198.0    Max.   :2654    Max.   :1299.0    Max.   :646.0
##      kms      PetrolPrice      VanKilled      law
## Min.   : 7685    Min.   :0.08118    Min.   : 2.000    Min.   :0.0000
## 1st Qu.:12685    1st Qu.:0.09258    1st Qu.: 6.000    1st Qu.:0.0000
## Median :14987    Median :0.10448    Median : 8.000    Median :0.0000
## Mean   :14994    Mean   :0.10362    Mean   : 9.057    Mean   :0.1198
## 3rd Qu.:17203    3rd Qu.:0.11406    3rd Qu.:12.000    3rd Qu.:0.0000
## Max.   :21626    Max.   :0.13303    Max.   :17.000    Max.   :1.0000
```

```
print('Dimensions of Seatbelts ')
```

```
## [1] "Dimensions of Seatbelts "
```

```
dim(Seatbelts)
```

```
## [1] 192    8
```

```
print('Rows in Seatbelts ')
```



```
## [1] "Rows in Seatbelts "
nrow(Seatbelts)
## [1] 192
print('Cols in Seatbelts ')
## [1] "Cols in Seatbelts "
ncol(Seatbelts)
## [1] 8
```

Seatbelts Dataset contains multiple time series with columns like Drivers killed , kilometers , petrol price, van killed. It contains 192 rows and 8 columns. Time series is from 1969 - 1985. It is processed dataset, there are no NaN values and also no null values in the dataset.

```
sum(is.na(Seatbelts))
## [1] 0
sum(is.null(Seatbelts))
## [1] 0
```

c. How is the built-in dataset `UKDriversDeaths` different from `Seatbelts` ?
#####Q1c#####

UKDriverDeaths

```
##      Jan  Feb  Mar  Apr  May  Jun  Jul  Aug  Sep  Oct  Nov  Dec
## 1969 1687 1508 1507 1385 1632 1511 1559 1630 1579 1653 2152 2148
## 1970 1752 1765 1717 1558 1575 1520 1805 1800 1719 2008 2242 2478
## 1971 2030 1655 1693 1623 1805 1746 1795 1926 1619 1992 2233 2192
## 1972 2080 1768 1835 1569 1976 1853 1965 1689 1778 1976 2397 2654
## 1973 2097 1963 1677 1941 2003 1813 2012 1912 2084 2080 2118 2150
## 1974 1608 1503 1548 1382 1731 1798 1779 1887 2004 2077 2092 2051
## 1975 1577 1356 1652 1382 1519 1421 1442 1543 1656 1561 1905 2199
## 1976 1473 1655 1407 1395 1530 1309 1526 1327 1627 1748 1958 2274
## 1977 1648 1401 1411 1403 1394 1520 1528 1643 1515 1685 2000 2215
## 1978 1956 1462 1563 1459 1446 1622 1657 1638 1643 1683 2050 2262
## 1979 1813 1445 1762 1461 1556 1431 1427 1554 1645 1653 2016 2207
## 1980 1665 1361 1506 1360 1453 1522 1460 1552 1548 1827 1737 1941
## 1981 1474 1458 1542 1404 1522 1385 1641 1510 1681 1938 1868 1726
## 1982 1456 1445 1456 1365 1487 1558 1488 1684 1594 1850 1998 2079
## 1983 1494 1057 1218 1168 1236 1076 1174 1139 1427 1487 1483 1513
## 1984 1357 1165 1282 1110 1297 1185 1222 1284 1444 1575 1737 1763
```

```
str(UKDriverDeaths)
```

```
## Time-Series [1:192] from 1969 to 1985: 1687 1508 1507 1385 1632 ...
```

```
print('Summary of UKDriverDeaths ')
```

```
## [1] "Summary of UKDriverDeaths "
```

```
summary(UKDriverDeaths)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      1057    1462    1631    1670    1851    2654
```

Datasets UKDriverDeaths & Seatbelts both are Timeseries data but UKDriverDeaths just have the count of deaths of driver with no other variables whereas in Seatbelts we have more variables in addition to the year and month data. UKDriverDeaths is a time series giving the monthly totals of car drivers in Great Britain killed or seriously injured Jan 1969 to Dec 1984. Compulsory wearing of seat belts was introduced on 31 Jan 1983. Seatbelts is more information on the same problem.

- d. What does Seatbelts[1,1] return? What does Seatbelts[29, 5] return? Describe in 1 sentence what is going on.

```
#####Q1d#####
```

```
Seatbelts[1,1]
```

```
## DriversKilled
```

```
##           107
```

```
Seatbelts[29, 5]
```

```
##      kms
```

```
## 13587
```

Seatbelts[1,1] returns the value(107) in cell[1,1] i.e. 1st row and 1st column(DriversKilled) of the dataframe & Seatbelts[29, 5] returns the value(13587) in 29th row and 5th column(kms) of the dataframe.

- e. If you were interested in analyzing deaths due to car accidents in the UK, describe how you could combine Seatbelts and UKDriversDeath to do so. (You do not need actually do this.)

What I can do is combine datasets and grab the deaths from UKdriverDeaths and combine it with the Front , vankilled over the years of 1969-1985. this will give the deaths in UK with van killed.

- f. Create a variable Bob and set it to True. Type Bob and what does R return. (Note that R is case sensitive so `Bob` does not equal `bob`.) Type `Bob+Bob`. What is the result and what is going on?

```
#####Q1f#####
```

```
Bob = TRUE
```

```
Bob+Bob
```

```
## [1] 2
```

Bob is assigned as 'TRUE' which is treated as boolean which is '1' for 'TRUE' & '0' for 'FALSE'. So, when we perform `Bob+Bob` it returns "2" which is 1+1 in boolean.

- g. What is vector recycling in R? (It applies to all vectors not just logical ones.) Create a vector of two logical values and another of 5 logical values. Ask R if those two vectors are equal. What happens and what is going on?

```
#####Q1g#####
```

```
a = c(1,0)
```

```
b= c(1,1,1,0,1)
```

```
a == b
```

```
## Warning in a == b: longer object length is not a multiple of shorter  
object
```

```
## length
```

```
## [1] TRUE FALSE TRUE TRUE TRUE
```

Vector Recycling in r: If two vectors are of unequal length, the shorter one will be recycled in order to match the longer vector. For example, the following vectors `u` and `v` have different lengths, and their sum is computed by recycling values of the shorter vector `u`.

if the value of "a" matches with Value of 'b' then R returns 'TRUE' otherwise 'FALSE'. Each value of vector 'a' is compared with vector 'b'. since 'a' has only 2 elements so for comparing the rest of the elements from 'b', 'a' is repeated. which means that R compares `a(1,0,1,0,1)` & `b(1,1,1,0,1)` which returns "TRUE FALSE TRUE TRUE TRUE"

Question 2: Basic Data Manipulation

- a. Download the PUMS dataset from Canvas, file name: `psam_p34.csv`. This is the Public Use Micro Dataset, a subset from the ACS survey. You can also find on Canvas the definition of variables for this dataset (`PUMS_Data_Dictionary_2017`). Follow the directions below to import a dataset into RStudio
- Go to `File` tab at top of Computer Screen

- Under `Import Dataset`, choose `From Text(base)`
- Navigate to the folder in which dataset is downloaded
- Click Import to continue through with the dataset
- **Note:** R Studio actually provides you with the code to import datasets. Type that code below

```
#####Q2a#####
```

```
psam <- read.csv("D:/Downloads/psam_p34.csv", stringsAsFactors=FALSE)
```

- b. Add a new column to the data frame and fill it with 10 in all rows.

```
#####Q2b#####
```

```
psam$newcol = 10
```

```
colnames(psam)
```

```
psam['newcol']
```

- c. Add a new column to the data frame and copy the data from an existing column in the dataset, PWGTP80 into this column.

```
#####Q2c#####
```

```
psam$newcol = psam$PWGTP80
```

```
psam['newcol']
```

- d. Delete the column PWGTP74

```
#####Q2d#####
```

```
psam$PWGTP74 <- NULL
```

```
#colnames(psam)
```

- e. Rename the column CIT as CHARCT

```
#####Q2e#####
```

```
names(psam)[names(psam) == 'CIT'] <- 'CHARCT'
```

```
#colnames(psam)
```

```
psam["CHARCT"]
```

Question 3: Subsetting & Sorting Data

Subsetting data is the process of retrieving just the parts of larger datasets that are of specific interest for the project at hand. It is a very important component of data management and there are several ways that one can subset data in R.

a. Complete the data subsetting tutorial at this [website](#)

b. Sort the data according to the variable MAR in ascending order

```
#####Q3b#####
```

```
head(psam)
```

```
psam1 = psam[order(psam$MAR),]
```

```
head(psam1)
```

c. Sort the data in ascending order by PWGTP3 and descending order by PWGTP7 together.

```
#####Q3c#####
```

```
psam2 = psam[order(psam$PWGTP3, -psam$PWGTP7),]
```

```
head(psam2)
```

d. Create a subset of the data by â keepingâ the first 10 variables in the PUMS dataset (RT to AGEP) or â droppingâ the other variables.

```
#####Q3d#####
```

```
psam_subset = psam[,1:10]
```

```
head(psam_subset)
```

```
dim(psam_subset)
```

e. Create a subset of the data by â keepingâ the first 10 observations.

```
#####Q3e#####
```

```
psam_sub = psam[1:10,]
```

```
head(psam_sub)
```

```
dim(psam_sub)
```

f. Take a random sample of the dataset of size 50:

– with replacement

```
#####Q3f#####
```

```
 #(i) with replacement
```

```
set.seed(100)
```

```
sam_rep = psam[sample(nrow(psam),50, replace = TRUE),]
```

```
head(sam_rep)
```

```
* without replacement
```

```
 #(ii) without replacement
```

```
sam_wrep = psam[sample(nrow(psam),50, replace = FALSE),]
```

```
head(sam_wrep)
```

Question 4: Descriptive Stats

For this question, we will use one of the most common R built-in datasets, `mtcars`. The easiest way to get descriptive statistics in R is using the `summary()` command.

a. Find the summary statistics of the `mtcars` dataset using the `summary()` command.

```
#####Q4a#####
```

```
summary(mtcars)
```

##	mpg	cyl	disp	hp
##	Min. :10.40	Min. :4.000	Min. : 71.1	Min. : 52.0
##	1st Qu.:15.43	1st Qu.:4.000	1st Qu.:120.8	1st Qu.: 96.5
##	Median :19.20	Median :6.000	Median :196.3	Median :123.0
##	Mean :20.09	Mean :6.188	Mean :230.7	Mean :146.7
##	3rd Qu.:22.80	3rd Qu.:8.000	3rd Qu.:326.0	3rd Qu.:180.0
##	Max. :33.90	Max. :8.000	Max. :472.0	Max. :335.0
##	drat	wt	qsec	vs
##	Min. :2.760	Min. :1.513	Min. :14.50	Min. :0.0000
##	1st Qu.:3.080	1st Qu.:2.581	1st Qu.:16.89	1st Qu.:0.0000

```
## Median :3.695   Median :3.325   Median :17.71   Median :0.0000
## Mean    :3.597   Mean    :3.217   Mean    :17.85   Mean    :0.4375
## 3rd Qu.:3.920   3rd Qu.:3.610   3rd Qu.:18.90   3rd Qu.:1.0000
## Max.    :4.930   Max.    :5.424   Max.    :22.90   Max.    :1.0000
##      am      gear      carb
## Min.    :0.0000   Min.    :3.000   Min.    :1.000
## 1st Qu.:0.0000   1st Qu.:3.000   1st Qu.:2.000
## Median :0.0000   Median :4.000   Median :2.000
## Mean    :0.4062   Mean    :3.688   Mean    :2.812
## 3rd Qu.:1.0000   3rd Qu.:4.000   3rd Qu.:4.000
## Max.    :1.0000   Max.    :5.000   Max.    :8.000
```

b. Another way to get more detailed descriptive statistics is to use the pastecs package.

- Install the pastecs package:

- Type `install.packages("pastecs")` and load it from the library by typing `library(pastecs)`

```
#####Q4b#####
```

```
#install.packages("pastecs")
library(pastecs)
```

```
## Warning: package 'pastecs' was built under R version 3.5.2
```

```
* Find the command to get the descriptive statistics using this package.
(Hint: your output should give you a minimum, maximum, range, SE. mean, C.I
Mean, standard deviation and coefficient of variance etc)
```

```
stat.desc(mtcars)
```

```
##      mpg      cyl      disp      hp
## nbr.val  32.0000000 32.0000000 3.200000e+01 32.0000000
## nbr.null  0.0000000 0.0000000 0.000000e+00  0.0000000
## nbr.na    0.0000000 0.0000000 0.000000e+00  0.0000000
## min      10.4000000  4.0000000 7.110000e+01 52.0000000
## max      33.9000000  8.0000000 4.720000e+02 335.0000000
## range    23.5000000  4.0000000 4.009000e+02 283.0000000
## sum      642.9000000 198.0000000 7.383100e+03 4694.0000000
## median   19.2000000  6.0000000 1.963000e+02 123.0000000
## mean     20.0906250  6.1875000 2.307219e+02 146.6875000
## SE.mean   1.0654240  0.3157093 2.190947e+01 12.1203173
## CI.mean.0.95 2.1729465 0.6438934 4.468466e+01 24.7195501
## var      36.3241028  3.1895161 1.536080e+04 4700.8669355
## std.dev   6.0269481  1.7859216 1.239387e+02 68.5628685
## coef.var  0.2999881  0.2886338 5.371779e-01  0.4674077
##      drat      wt      qsec      vs      am
## nbr.val 32.00000000 32.0000000 32.0000000 32.00000000 32.00000000
## nbr.null 0.00000000 0.0000000 0.0000000 18.00000000 19.00000000
## nbr.na  0.00000000 0.0000000 0.0000000 0.00000000 0.00000000
```

```
## min          2.76000000  1.5130000  14.5000000  0.00000000  0.00000000
## max          4.93000000  5.4240000  22.9000000  1.00000000  1.00000000
## range        2.17000000  3.9110000   8.4000000  1.00000000  1.00000000
## sum          115.09000000 102.9520000 571.1600000 14.00000000 13.00000000
## median       3.69500000  3.3250000  17.7100000  0.00000000  0.00000000
## mean        3.59656250  3.2172500  17.8487500  0.43750000  0.40625000
## SE.mean      0.09451874  0.1729685   0.3158899  0.08909831  0.08820997
## CI.mean.0.95 0.19277224  0.3527715   0.6442617  0.18171719  0.17990541
## var          0.28588135  0.9573790   3.1931661  0.25403226  0.24899194
## std.dev      0.53467874  0.9784574   1.7869432  0.50401613  0.49899092
## coef.var      0.14866382  0.3041285   0.1001159  1.15203687  1.22828533
##              gear      carb
## nbr.val      32.0000000 32.0000000
## nbr.null      0.0000000 0.0000000
## nbr.na        0.0000000 0.0000000
## min          3.0000000 1.0000000
## max          5.0000000 8.0000000
## range        2.0000000 7.0000000
## sum          118.0000000 90.0000000
## median       4.0000000 2.0000000
## mean        3.6875000 2.8125000
## SE.mean      0.1304266 0.2855297
## CI.mean.0.95 0.2660067 0.5823417
## var          0.5443548 2.6088710
## std.dev      0.7378041 1.6152000
## coef.var      0.2000825 0.5742933
```

- c. There are also separate commands to get the mean, median and mean statistics. Find the mean, median and mode of the variable mpg by separate commands.

```
#####Q4c#####

mean(mtcars$mpg)

## [1] 20.09062

median(mtcars$mpg)

## [1] 19.2

mode1 = names(table(mtcars$mpg))[table(mtcars$mpg)==max(table(mtcars$mpg))]

mode1

## [1] "10.4" "15.2" "19.2" "21"   "21.4" "22.8" "30.4"
```

- d. Find the length of the variable qtsec. Why are the lengths of all the variables the same?


```
#####Q4d#####
```

```
length(mtcars$qsec)
```

```
## [1] 32
```

```
# Length of 'qsec' is 32
```

```
# Variables are part of dataframe , which is a table and each variable needs  
to be equal length to form a tabular structure. even if variables are of  
unequal length they are filled with NULL to make up the dimensions of table.
```

e. Find the maximum and minimum value of the mpg variable.

```
#####Q4e#####
```

```
mpg_max = max(mtcars$mpg)  
mpg_max
```

```
## [1] 33.9
```

```
mpg_min= min(mtcars$mpg)  
mpg_min
```

```
## [1] 10.4
```

f. Determine the location i.e index of the maximum and minimum value you found in part e. (Hint: Try the which.max command).

```
#####Q4f#####
```

```
which.max(mtcars$mpg)
```

```
## [1] 20
```

```
which.min(mtcars$mpg)
```

```
## [1] 15
```

Question 5: Putting it all together

Downloading a dataset:

1. Go to [this link](#) on Kaggle. This should take you to a page for the “San Francisco Building Permits” dataset. (**note:** you will have to create an account in order to download this dataset. Kaggle is a PHENOMENAL resource for datasets and data-related explorations, so making this account now will help you for future assignments.)

2. Once youâ€™ve downloaded this dataset, time to **import** the dataset into RStudio (Refer to Question 2 for tips on importing datasets). Type in the code that imports this dataset below: (After a while, you will see that the dataset â€™Building_Permitsâ€™ is available in your â€™Global Variableâ€™ explorer in RStudio)

```
#####Q5->2#####
```

```
Building_Permits <- read.csv("C:/Users/akhil/Downloads/Building_Permits.csv",  
stringsAsFactors=FALSE)
```

Preparing for data manipulation:

1. If youâ€™ve successfully imported the dataset, you should have a Building_Permits variable in your global explorer – **congrats!** As per convention, itâ€™s always a great idea to create a **copy** of your dataset, so that whatever manipulations you make donâ€™t affect the original dataset. With that being said, make a copy of the dataset! *
hint: Name the copy whatever you would like and literally use the <- operator to assign your newly named variable to the existing Building_Permits dataset

```
#####Q5->2->1#####
```

```
buldgp = Building_Permits
```

Exploring the dataset:

1. Thus far, weâ€™ve downloaded the dataset and made copies to prevent against any future accidents. Now, letâ€™s explore our dataset a little further and really understand what weâ€™re dealing with here:
2. Reproduce the following printed statement **with code** and **replace X** with the number of rows and **replace Y** with the number of columns of your dataset: Dimensions: X rows, Y columns

- **hints:**

1. youâ€™ll find the [R cat\(\)](#) function really helpful
2. The dim() function from question 1 will be really useful! (Also, there are two components that are returned by calling the dim() function, and you can access each portion with a proper index call (example:
dim(â€™<dataset_nameâ€™)[index])
3. You can use the cat() function as follows: cat(â€™<String: â€™ , data, â€™<another stringâ€™ , more data)

```
#####Q5->3->2#####
```

```
dim(buldgp)
```

```
cat("Dimensions: ",dim(buldgp)[1], " rows, ",dim(buldgp)[2], " columns" )
```

3. Generally, if weâre dealing with data that is numeric, it might be helpful to look for the averages in a dataset. Take a quick look at the different columns of this dataset – do you think itâs appropriate to analyze stats like the mean, median, mode for the numeric columns? Why or why not?

```
#####Q5->3->3#####
```

```
str(bulldgp)
```

```
## 'data.frame': 198900 obs. of 43 variables:
## $ Permit.Number : chr "201505065519"
"201604195146" "201605278609" "201611072166" ...
## $ Permit.Type : int 4 4 3 8 6 8 8 8 8 8 ...
## $ Permit.Type.Definition : chr "sign - erect" "sign -
erect" "additions alterations or repairs" "otc alterations permit" ...
## $ Permit.Creation.Date : chr "05/06/2015" "04/19/2016"
"05/27/2016" "11/07/2016" ...
## $ Block : chr "0326" "0306" "0595"
"0156" ...
## $ Lot : chr "023" "007" "203" "011"
...
## $ Street.Number : int 140 440 1647 1230 950 800
1291 1465 2094 89 ...
## $ Street.Number.Suffix : chr "" "" "" "" ...
## $ Street.Name : chr "Ellis" "Geary" "Pacific"
"Pacific" ...
## $ Street.Suffix : chr "St" "St" "Av" "Av" ...
## $ Unit : int NA 0 NA 0 NA NA 0 NA NA NA
...
## $ Unit.Suffix : chr "" "" "" "" ...
## $ Description : chr "ground fl facade: to
erect illuminated, electric, wall, single faced sign. n/a for maher ordinance
155-13." "remove (e) awning and associated signs." "installation of
separating wall" "repair dryrot & stucco at front of bldg." ...
## $ Current.Status : chr "expired" "issued"
"withdrawn" "complete" ...
## $ Current.Status.Date : chr "12/21/2017" "08/03/2017"
"09/26/2017" "07/24/2017" ...
## $ Filed.Date : chr "05/06/2015" "04/19/2016"
"05/27/2016" "11/07/2016" ...
## $ Issued.Date : chr "11/09/2015" "08/03/2017"
"" "07/18/2017" ...
## $ Completed.Date : chr "" "" "" "07/24/2017" ...
## $ First.Construction.Document.Date : chr "11/09/2015" "08/03/2017"
"" "07/18/2017" ...
## $ Structural.Notification : chr "" "" "" "" ...
## $ Number.of.Existing.Stories : num 6 7 6 2 3 5 3 NA NA NA ...
## $ Number.of.Proposed.Stories : num NA NA 6 2 NA 5 3 NA NA NA
...
## $ Voluntary.Soft.Story.Retrofit : chr "" "" "" "" ...
## $ Fire.Only.Permit : chr "" "" "" "" ...
## $ Permit.Expiration.Date : chr "11/03/2016" "12/03/2017"
```

```

"" "07/13/2018" ...
## $ Estimated.Cost : num 4000 1 20000 2000 100000
4000 12000 NA NA NA ...
## $ Revised.Cost : num 4000 500 NA 2000 100000
4000 12000 0 1 0 ...
## $ Existing.Use : chr "tourist hotel/motel"
"tourist hotel/motel" "retail sales" "1 family dwelling" ...
## $ Existing.Units : num 143 NA 39 1 NA 326 5 NA NA
NA ...
## $ Proposed.Use : chr "" "" "retail sales" "1
family dwelling" ...
## $ Proposed.Units : int NA NA 39 1 NA 326 5 NA NA
NA ...
## $ Plansets : int 2 2 2 2 2 2 0 NA NA NA ...
## $ TIDF.Compliance : chr "" "" "" "" ...
## $ Existing.Construction.Type : int 3 3 1 5 3 1 5 NA NA NA ...
## $ Existing.Construction.Type.Description: chr "constr type 3" "constr
type 3" "constr type 1" "wood frame (5)" ...
## $ Proposed.Construction.Type : int NA NA 1 5 NA 1 5 NA NA NA
...
## $ Proposed.Construction.Type.Description: chr "" "" "constr type 1"
"wood frame (5)" ...
## $ Site.Permit : chr "" "" "" "" ...
## $ Supervisor.District : int 3 3 3 3 6 10 5 10 5 8 ...
## $ Neighborhoods...Analysis.Boundaries : chr "Tenderloin" "Tenderloin"
"Russian Hill" "Nob Hill" ...
## $ Zipcode : int 94102 94102 94109 94109
94102 94107 94122 94124 94117 94117 ...
## $ Location : chr "(37.785719256680785, -
122.40852313194863)" "(37.78733980600732, -122.41063199757738)"
"(37.7946573324287, -122.42232562979227)" "(37.79595867909168, -
122.41557405519474)" ...
## $ Record.ID : num 1.38e+12 1.42e+12 1.42e+12
1.44e+12 1.45e+11 ...

```

summary(bulldgp)

```

## Permit.Number      Permit.Type      Permit.Type.Definition
## Length:198900      Min.      :1.000      Length:198900
## Class :character    1st Qu.:8.000      Class :character
## Mode  :character    Median :8.000      Mode  :character
##                      Mean  :7.522
##                      3rd Qu.:8.000
##                      Max.  :8.000
##
## Permit.Creation.Date  Block      Lot      Street.Number
## Length:198900        Length:198900  Length:198900  Min.      : 0
## Class :character     Class :character  Class :character  1st Qu.: 235
## Mode  :character     Mode  :character  Mode  :character  Median   : 710
##                      Mean    :1122

```

```

##                                     3rd Qu.:1700
##                                     Max.    :8400
##
## Street.Number.Suffix Street.Name      Street.Suffix
## Length:198900      Length:198900      Length:198900
## Class :character    Class :character    Class :character
## Mode  :character    Mode  :character    Mode  :character
##
##
##
##      Unit          Unit.Suffix      Description
## Min.   : 0.00      Length:198900      Length:198900
## 1st Qu.: 0.00      Class :character    Class :character
## Median : 0.00      Mode  :character    Mode  :character
## Mean   : 78.52
## 3rd Qu.: 1.00
## Max.   :6004.00
## NA's   :169421
## Current.Status      Current.Status.Date Filed.Date
## Length:198900      Length:198900      Length:198900
## Class :character    Class :character    Class :character
## Mode  :character    Mode  :character    Mode  :character
##
##
##
## Issued.Date          Completed.Date      First.Construction.Document.Date
## Length:198900      Length:198900      Length:198900
## Class :character    Class :character    Class :character
## Mode  :character    Mode  :character    Mode  :character
##
##
##
## Structural.Notification Number.of.Existing.Stories
## Length:198900      Min.    : 0.00
## Class :character    1st Qu.: 2.00
## Mode  :character    Median : 3.00
##                      Mean    : 5.71
##                      3rd Qu.: 4.00
##                      Max.    :78.00
##                      NA's    :42784
## Number.of.Proposed.Stories Voluntary.Soft.Story.Retrofit
## Min.    : 0.00      Length:198900
## 1st Qu.: 2.00      Class :character
## Median : 3.00      Mode  :character
## Mean    : 5.75
## 3rd Qu.: 4.00
## Max.    :78.00

```

```

## NA's :42868
## Fire.Only.Permit Permit.Expiration.Date Estimated.Cost
## Length:198900 Length:198900 Min. :1.00e+00
## Class :character Class :character 1st Qu.:3.30e+03
## Mode :character Mode :character Median :1.10e+04
## Mean :1.69e+05
## 3rd Qu.:3.50e+04
## Max. :5.38e+08
## NA's :38066
## Revised.Cost Existing.Use Existing.Units
## Min. : 0 Length:198900 Min. : 0.00
## 1st Qu.: 1 Class :character 1st Qu.: 1.00
## Median : 7000 Mode :character Median : 1.00
## Mean : 132856 Mean : 15.67
## 3rd Qu.: 28708 3rd Qu.: 4.00
## Max. :780500000 Max. :1907.00
## NA's :6066 NA's :51538
## Proposed.Use Proposed.Units Plansets TIDF.Compliance
## Length:198900 Min. : 0.00 Min. : 0.00 Length:198900
## Class :character 1st Qu.: 1.00 1st Qu.: 0.00 Class :character
## Mode :character Median : 2.00 Median : 2.00 Mode :character
## Mean : 16.51 Mean : 1.27
## 3rd Qu.: 4.00 3rd Qu.: 2.00
## Max. :1911.00 Max. :9000.00
## NA's :50911 NA's :37309
## Existing.Construction.Type Existing.Construction.Type.Description
## Min. :1.00 Length:198900
## 1st Qu.:3.00 Class :character
## Median :5.00 Mode :character
## Mean :4.07
## 3rd Qu.:5.00
## Max. :5.00
## NA's :43366
## Proposed.Construction.Type Proposed.Construction.Type.Description
## Min. :1.00 Length:198900
## 1st Qu.:3.00 Class :character
## Median :5.00 Mode :character
## Mean :4.09
## 3rd Qu.:5.00
## Max. :5.00
## NA's :43162
## Site.Permit Supervisor.District
## Length:198900 Min. : 1.000
## Class :character 1st Qu.: 3.000
## Mode :character Median : 6.000
## Mean : 5.538
## 3rd Qu.: 8.000
## Max. :11.000
## NA's :1717
## Neighborhoods...Analysis.Boundaries Zipcode Location

```

```
## Length:198900           Min.    :94102   Length:198900
## Class :character        1st Qu.:94109   Class :character
## Mode  :character        Median :94114   Mode  :character
##                          Mean    :94116
##                          3rd Qu.:94122
##                          Max.    :94158
##                          NA's    :1716
## Record.ID
## Min.    :1.294e+10
## 1st Qu.:1.309e+12
## Median :1.372e+12
## Mean    :1.162e+12
## 3rd Qu.:1.435e+12
## Max.    :1.498e+12
##
```

*# No, Because by running the above commands we can see that most of the columns in the dataset are character strings.
but there are some columns like units , Estimated cost , revised cost , etc. for which we can have need for mean , median.*

4. We know that weâre dealing with an incredible number of rows in our dataset (if you discovered the dimensions properly, weâre looking at ~200k rows). However, for some columns, we donât have ~200k unique values. Letâs discover some unique values. Find the number of unique values that are in the Existing.Use and the Neighborhoods...Analysis.Boundaries columns and print your results in the following format, replacing X and Y with their appropriate values (Please donât just *write* in the numbers, we want to see you **use** the functions in R to figure this out!) The Existing.Use column has X unique values and the Neighborhoods...Analysis.Boundaries has Y unique values

- **hints:**

1. `cat()` will be your best friend!
2. There is literally a function called `unique()`– figure out how to manipulate this!

```
#####Q5->3->4#####
#colnames(buldgp)
```

```
cat("The Existing.Use column has ", length(unique(buldgp$Existing.Use)), "
unique values and the Neighborhoods...Analysis.Boundaries has
",length(unique(buldgp$Neighborhoods...Analysis.Boundaries)) ," unique
values")
```

5. This is the DIY part of your data exploration – find something interesting about the data using R code, and tell us why you think itâs interesting!

```
colnames(buldgp)
library(ggplot2)
```

```
# making a copy of dataset to perent changes.
temp = buldgp
```

```

str(temp)

#parsing the FiledDate column to recognize as Date in R
buldgp$Filed.Date <- as.Date(buldgp$Filed.Date, "%m/%d/%Y")

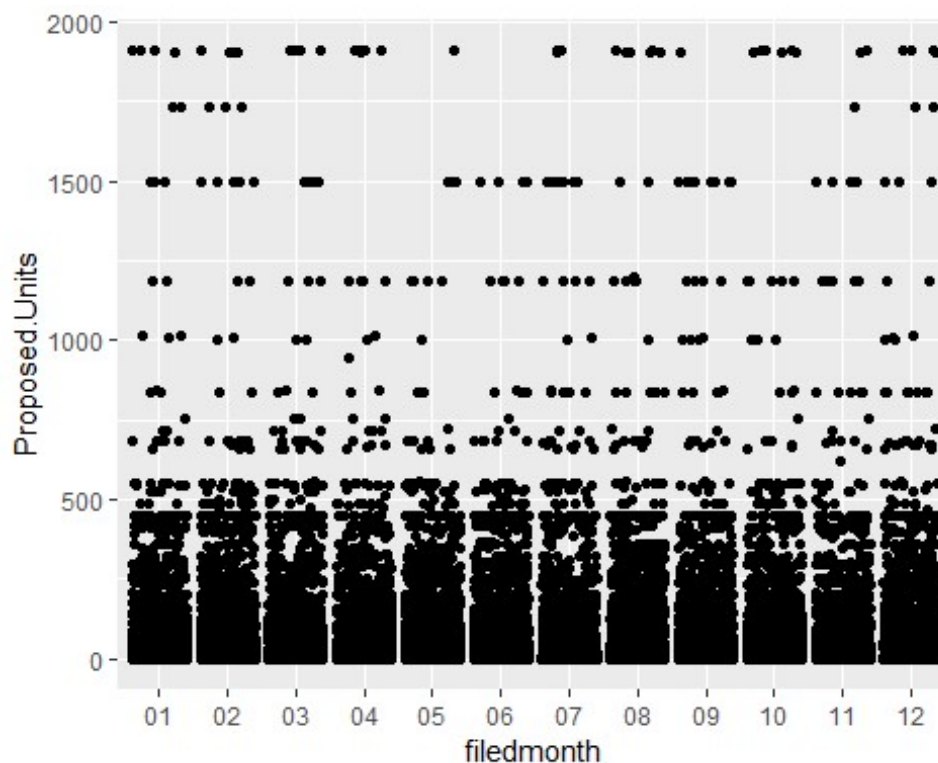
#creating new column with filedmonth
temp$filedmonth = format(buldgp$Filed.Date, '%m')

#creating new column with filedeyear
temp$filedeyear = format(buldgp$Filed.Date, '%Y')

# plotting filedmonth with the proposed.Units
ggplot(temp,aes(x=filedmonth , y = Proposed.Units)) +
  geom_jitter()

## Warning: Removed 50911 rows containing missing values (geom_point).

```

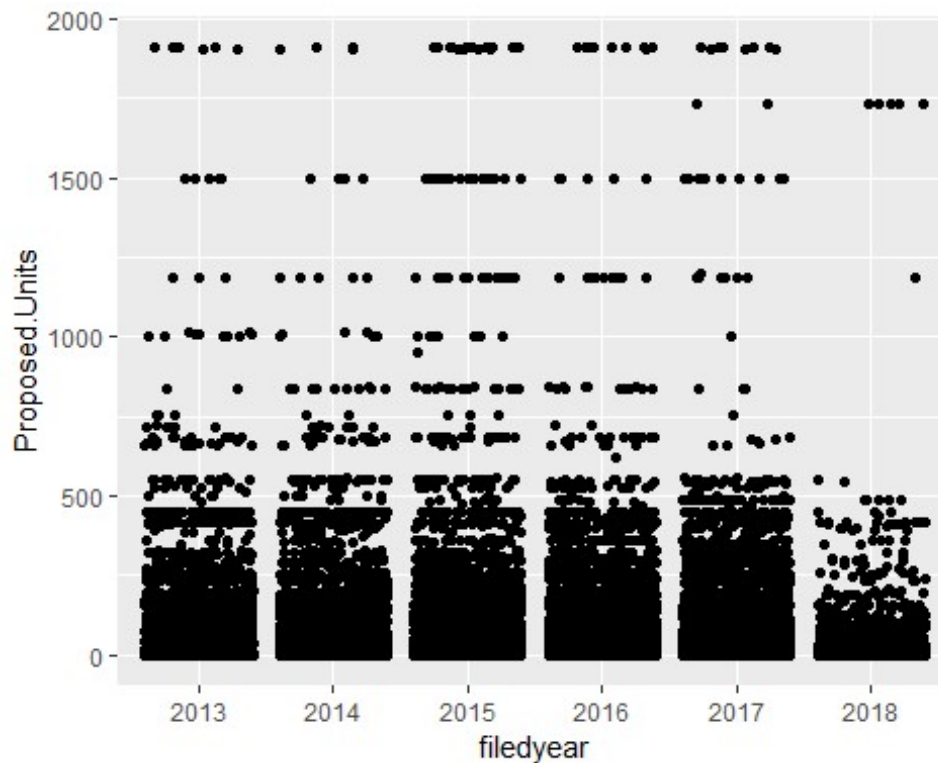


```

# plotting filedeyear with the proposed.Units
ggplot(temp,aes(x=filedeyear , y = Proposed.Units)) +
  geom_jitter()

## Warning: Removed 50911 rows containing missing values (geom_point).

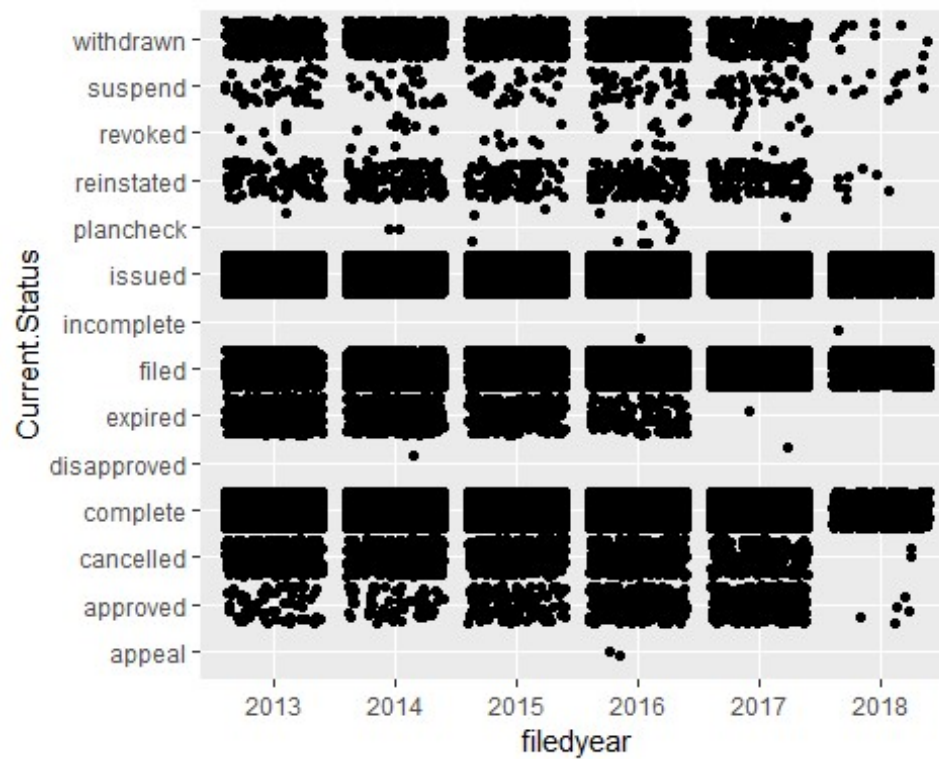
```

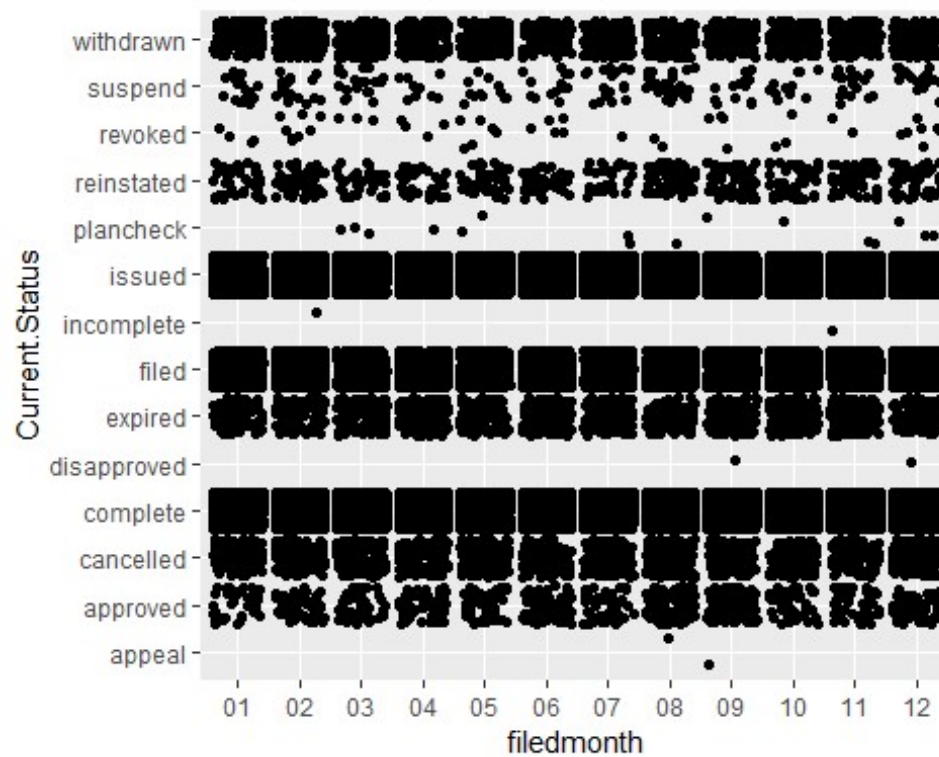
From the above graph we can conclude that most of the proposed units were in the months of January & May and the highest number of proposed units is 1500 in months of January. Also, Looking at the year graph we can say that most of the units were proposed in 2013 and some of units were proposed in 2017.

```
temp$Current.Status = as.factor( temp$Current.Status)
```

```
ggplot(temp,aes(x=filedyear , y = Current.Status)) +  
  geom_jitter()
```



```
ggplot(temp,aes(x=filedmonth , y = Current.Status)) +  
  geom_jitter()
```



From the above graph we can conclude that Current status of most of the Building permits were in 2013. Also in 2013, the Current Status of permits were mostly issued, filed, complete & somewhat withdrawn. Some of the permits were issued in 2017. From the other graph current status of projects are mostly concentrated in first 5 months and then permits are taken down in rest of the year. Most of the permits appear to be issued & completed till the month of May.

Data Manipulation

Let's draw upon Question 3: Subsetting datasets. Oftentimes, when we're working with data, we're not concerned about every single column in a dataset. Instead, there is only a handful of columns that are important to our needs. With this in mind, we'll subset our dataset so that we don't have to continually sift through relatively useless information in order to use our data. To this effect, we're going to create 2 individual datasets that are simply subsets of our main, overarching dataset.

- Create a subset of your **copy** of the Building_Permits dataset that only contains the following columns: Permit.Number, Description, Existing.Use

```
#####Q5->4->1a#####
```

```
#colnames(buldp)
```

```
#buldp_sub =  
cbind(buldp$Existing.Use, buldp$Permit.Number, buldp$Description)
```

```
buldp_sub1 = buldp[c("Existing.Use", "Description", "Permit.Number")]
```

```
dim(buldp_sub1)
```

- Create a second subset of your **copy** of the Building_Permits dataset that only contains the following columns: Permit.Number, Proposed.Use. However, we want this subset to only access entries from row 50,000 to 60,000.
 - hints:**
 - when subsetting for specific entries in a dataset, we can actually do the following: `dataset[index, index][<condition>]`
 - To access rows in a column, we specify the index to be `dataset[index,]`. The lefthand side is for specifying rows, the righthand side is for specifying columns

```
#####Q5->4->b#####
```

```
buldp_sub2 = buldp[50000:60000, c("Permit.Number", "Proposed.Use")]
```

- Now that we have two separate components of our dataset, let's merge them together! Realistically, you'd really just create a singular subset with this information together. However, we have a highly specific use case now: one of our subsets only refers to a portion of the entries in our dataset, while the other dataset

refers to all of the entries in our dataset 1. Merging datasets requires a really, really longwinded and misleading complex function: `merge()` (this was a miserable joke by one of your TAs, feel free to send hate mail to Sridhar). Read the documentation, understand the parameters, and merge the datasets based on the `Permit.Number` column into a new variable.

```
#####Q5->4->c#####
```

```
buldg_sub = merge(buldg_sub1,buldg_sub2,by="Permit.Number")
```

```
head(buldg_sub,10)
```

```
dim(buldg_sub)
```

- d. There's now an interesting phenomenon regarding our dataset: even though the second subset dealt with rows 50,000 to 60,000 (~ 10k entries), our new dataset does not match the ~10k dimension! Why do you think this is? (**hint**: `unique()` might come in handy)

```
#####Q5->4->d#####
```

```
length(unique(buldg_sub$Permit.Number))
```

```
## [1] 9221
```

#By default the data frames are merged on the columns with names they both have, but separate specifications of the columns can be given by `by.x` and `by.y`. Columns can be specified by name, number or by a logical vector: the name "row.names" or the number 0 specifies the row names. The rows in the two data frames that match on the specified columns are extracted, and joined together. If there is more than one match, all possible matches contribute one row each.

#So, in our case the 10,000 values of second dataframe matched with the ~200k values in 1st data frame and if there were more than one match it led to their individual columns. which led to a total of 11783 values means we have unique values as 9221, so 2562 are the duplicates that matched more than one pair.

- e. Let's take this newly merged dataset, and alphabetize the data based on the `Proposed.Use` column. The `order()` function will help tremendously!

```
#####Q5->4->e#####
```

```
buldgp = buldgp[order(buldg_sub$Proposed.Use),]
```

```
head(buldgp)
```

f. Take a look at your new, alphabetized dataset. In the `Proposed.Use` column, we're missing data for what seems to be a decent amount of columns' entries. Normally, we'd use the `is.na()` or `is.null()` function like we did earlier to check for missing data. However, in this dataset, all empty data are actually considered to be *empty strings*. (Example: `" "`). It sounds really counterintuitive but despite these entries being visibly empty, R considers them to be non-empty entries. With this in mind, let's tackle the missing data:

1. Find the number of missing data points in the `Proposed.Use` column. (You'll need to check which entries are **empty strings**)

```
#####Q5->4->f#####
```

```
sum(buldg_sub$Proposed.Use == "")
```

```
which(buldg_sub$Proposed.Use == "")
```

g. Through a stroke of luck, Dr. Felder recently stumbled on a bit of cash and has decided to quit his job as a professor and invest in real estate full time! (again, a miserable joke). To help him with this, we want to replace all of the missing entries that we found in the `Proposed.Use` column with `"felder's penthouse"`

- **warning:** this is not an easy task and requires a bit of thinking. [This post](#) on StackOverflow is really insightful to approach this problem.
 - This post converts the existing column to a character datatype with `as.character()` because even though that we can see that the entries in a column are text, R sometimes encodes text-based columns as different data types. To guard against this, we use `as.character()`.
- **side note:** side note (optional): sometimes, we want to export datasets that we create so that others can use them! [write.csv\(\)](#) is a really helpful way to write any dataframes to .csv files!

```
#####Q5->4->g#####
```

```
buldg_sub$Proposed.Use[buldg_sub$Proposed.Use == ""] <- "felder's penthouse"
```

```
sum(buldg_sub$Proposed.Use == "felder's penthouse")
```

Feedback

- a. How long did it take to complete this homework? -> Almost 2 days
- b. How difficult was the homework? -> 6 (on scale of 1-10) it was not difficult just the thing is it had many questions. So, it took time .

- c. Which parts did you find useful and which parts were less useful? -> Q5 was challenging and whole assignment is usefull.
- d. What suggestions do you have regarding the lectures or homework assignments that would improve them? -> instead of putting questions in Rmd file you can just put chunks and write their respesctive question numbers so that it becomes easy to navigate and code looks much clean. whereas in this current scenario it becomes too crowded and much harder to navigate to any sub question. just put the questions in assignmnet pdf we can reffer question from their. example :

```
##### this is an example for Question number#####  
# this above line helps to navigate to respective chunks.
```