

MULTIPLE LINEAR REGRESSION

```
fit<- lm(rating~bpm+nrgy+dnce+dB+val+dur+acous+spch,data = data_clean)
summary(fit)
```

```
## Call:
## lm(formula = rating ~ bpm + nrgy + dnce + dB + val + dur + acous +
##      spch, data = data_clean)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.0684 -0.1588  0.1418  0.2854  1.4438
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  5.0729840  0.4549944  11.150 < 2e-16 ***
## bpm          0.0006058  0.0012819   0.473 0.636706
## nrgy         -0.0099612  0.0028328  -3.516 0.000471 ***
## dnce          0.0035088  0.0028024   1.252 0.211032
## dB           0.0601839  0.0136253   4.417 1.19e-05 ***
## val          -0.0003163  0.0017381  -0.182 0.855654
## dur          -0.0021127  0.0009286  -2.275 0.023259 *
## acous         -0.0003267  0.0018587  -0.176 0.860553
## spch         -0.0009379  0.0041604  -0.225 0.821725
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.7446 on 593 degrees of freedom
## Multiple R-squared:  0.059, Adjusted R-squared:  0.0463
## F-statistic: 4.647 on 8 and 593 DF, p-value: 1.587e-05
```

```
coefficients(fit)
```

```
##      (Intercept)          bpm          nrgy          dnce
## dB
##  5.0729840484  0.0006057746 -0.0099611954  0.0035088480  0.060183
## 8831
##          val          dur          acous          spch
## -0.0003163147 -0.0021126753 -0.0003266664 -0.0009378581
```

```
library(GGally)
```

```
confint(fit,level=0.95)
```

	2.5 %	97.5 %
## (Intercept)	4.179387646	5.9665804510
## bpm	-0.001911876	0.0031234256
## nrgy	-0.015524755	-0.0043976362
## dnce	-0.001994987	0.0090126835
## dB	0.033424245	0.0869435216
## val	-0.003729900	0.0030972710
## dur	-0.003936471	-0.0002888796
## acous	-0.003977183	0.0033238503
## spch	-0.009108726	0.0072330102

Predicted Values fitted(fit)

	1	2	3	4	5	6
## 7	8					
## 3.7458413	3.5562510	3.9336673	3.5840761	3.7410124	3.7272929	3.9278802
## 9	10	11	12	13	14	
## 15	16					
## 3.9599573	3.9447971	3.7455732	3.7350415	3.8236817	3.7568027	3.4592303
## 17	18	19	20	21	22	
## 23	24					
## 4.0930899	3.4764473	3.5405644	4.1597275	3.7643827	3.8629393	3.9461250
## 25	26	27	28	29	30	
## 31	32					
## 3.6237872	3.6681350	3.7081134	3.7420871	3.7957417	3.9242591	4.0914590
## 33	34	35	36	37	38	
## 39	40					
## 3.9978524	3.8013570	3.6389334	4.0607846	3.9748719	3.7158894	3.5288588
## 41	42	43	44	45	46	
## 47	48					
## 3.8010693	3.9713729	3.7581985	3.8667744	3.8534561	3.8907832	3.6742832
## 49	50	51	52	53	54	
## 55	56					
## 3.3694824	3.7688554	3.8079718	3.8547484	3.8982124	3.6989465	3.7410124
## 57	58	59	60	61	62	
## 63	64					
## 3.7330589	3.9560987	3.7142941	3.9207044	3.8357552	3.7008105	3.6476228
## 65	66	67	68	69	70	
## 71	72					
## 3.7350415	3.8842333	3.8567833	3.8298573	3.8255487	3.6204266	3.6983124
## 3.7015630						

##	73	74	75	76	77	78
79	80					
##	3.8201277	3.8065652	4.0537785	3.8244589	3.5326135	3.6875769
4394	3.9557991					
##	81	82	83	84	85	86
87	88					
##	3.9467763	3.6615055	3.6674119	3.7441728	3.8340883	3.9209860
5535	3.8142766					
##	89	90	91	92	93	94
95	96					
##	3.8632847	3.8846055	3.7778766	3.5967382	3.8734081	3.6742832
1438	3.3694824					
##	97	98	99	100	101	102
103	104					
##	3.8166924	3.7948420	3.8582821	3.8362624	3.5603378	3.8152403
4547	3.8493477					
##	105	106	107	108	109	110
111	112					
##	3.7934124	3.9531848	3.7064833	3.8414349	4.0399708	3.9576645
6294	4.0939373					
##	113	114	115	116	117	118
119	120					
##	3.9054028	3.8674362	3.6658156	3.7780806	4.0927214	3.8954148
9437	3.8759663					
##	121	122	123	124	125	126
127	128					
##	3.7193125	3.8384017	3.6121949	4.0977104	3.6885415	4.0498513
6167	3.8796093					
##	129	130	131	132	133	134
135	136					
##	3.8474728	3.7397592	3.6496004	3.7511013	3.7573180	3.8246690
5158	3.9075884					
##	137	138	139	140	141	142
143	144					
##	3.8966934	3.5332865	3.6872717	3.7286527	3.6492283	3.7992903
6493	3.7077722					
##	145	146	147	148	149	150
151	152					
##	3.8155375	3.8504370	3.8756763	3.5157365	4.0579933	3.8470462
2698	3.7927219					
##	153	154	155	156	157	158
159	160					
##	3.8674362	3.9235667	3.7858492	3.7173707	3.8251940	3.7719678
8049	3.8720266					
##	161	162	163	164	165	166
167	168					
##	3.8199337	3.6982050	3.5085653	3.5624128	3.8046059	3.6339691
3671	3.8480915					
##	169	170	171	172	173	174

175	176					
## 3.7294212	3.9063662	3.7520478	3.7670257	3.8530546	3.8526660	3.9506514
3.8890795						
##	177	178	179	180	181	182
183	184					
## 3.9190723	3.7396361	3.8968986	3.9004508	3.9021217	3.7415003	3.8058848
3.7493327						
##	185	186	187	188	189	190
191	192					
## 3.8338509	3.7953029	4.0910731	3.5390443	3.3378632	3.7732598	3.8122235
3.6864398						
##	193	194	195	196	197	198
199	200					
## 3.7984279	3.7148772	3.6139405	3.8985291	3.7301841	3.6930839	3.8471700
3.8684394						
##	201	202	203	204	205	206
207	208					
## 3.6843343	3.9106586	3.9525350	3.7016640	3.8418899	3.8285287	3.8034022
3.7904843						
##	209	210	211	212	213	214
215	216					
## 3.8359441	3.6831861	3.9990428	4.0975560	3.7656550	3.7335580	3.8625993
3.7898550						
##	217	218	219	220	221	222
223	224					
## 3.9601094	3.7057857	3.9042727	4.0473212	3.7777111	3.8146785	3.8337025
3.9416099						
##	225	226	227	228	229	230
231	232					
## 3.6198521	3.7203541	3.7886623	3.7896717	3.7319448	3.7019220	4.0478362
3.9088222						
##	233	234	235	236	237	238
239	240					
## 3.8970495	3.9192156	3.7803395	3.8728149	3.7212476	3.9129956	3.7163834
3.8968986						
##	241	242	243	244	245	246
247	248					
## 3.8973584	3.4615230	3.9433286	3.7458248	3.9630282	3.8245888	3.7352349
3.8773525						
##	249	250	251	252	253	254
255	256					
## 3.8929193	4.1055232	3.7288297	3.8424493	3.8394528	3.5193741	3.6891726
4.1016995						
##	257	258	259	260	261	262
263	264					
## 3.6258706	3.9407850	3.9345245	3.8662573	4.0912938	3.9184495	3.8786008
3.9886115						
##	265	266	267	268	269	270
271	272					

3.8502434 3.7683629 3.7924462 3.6528629 3.9525694 3.9705638 3.843
1639 3.7877169
273 274 275 276 277 278
279 280
3.8078198 3.8423205 3.6715310 3.9189310 4.3295413 3.6941716 4.075
7667 3.9088469
281 282 283 284 285 286
287 288
3.7900550 3.9110009 3.9394507 3.9294868 3.8581909 3.8305576 3.813
1238 3.8405680
289 290 291 292 293 294
295 296
3.6811901 3.7274339 3.9301773 3.9275442 3.8872622 3.6474633 3.909
9275 3.8814315
297 298 299 300 301 302
303 304
3.8293126 3.9280237 3.7483845 3.9212899 3.6712614 3.7768278 3.984
5654 3.9288647
305 306 307 308 309 310
311 312
3.9615003 3.9315080 3.6278194 3.8569254 3.7401873 3.8260153 3.601
3051 4.0378093
313 314 315 316 317 318
319 320
3.8032951 3.8801861 3.8506137 3.7786070 3.6918065 3.6222984 3.691
6309 3.6354377
321 322 323 324 325 326
327 328
3.8005055 3.8567342 3.6715310 3.6288739 3.8850693 3.7973767 4.050
4516 3.7678750
329 330 331 332 333 334
335 336
3.7522044 3.9136777 4.1436255 3.6794350 3.4475733 3.9643792 3.858
9009 3.7302063
337 338 339 340 341 342
343 344
3.7651226 3.8081710 3.7992659 3.8895864 3.8822799 3.7399900 3.735
8684 3.8050913
345 346 347 348 349 350
351 352
3.8060340 3.6739291 3.8099737 4.0866493 3.8768868 3.6666133 3.918
1484 3.7029429
353 354 355 356 357 358
359 360
3.6392233 3.8071149 3.7275938 3.8427557 3.8356079 3.7354687 3.964
7426 3.8514848
361 362 363 364 365 366
367 368
3.7130535 4.0619651 4.0684367 3.8431639 3.7877169 3.8665100 3.714

4416 4.0237083
369 370 371 372 373 374
375 376
3.7295195 3.7689871 4.0293851 4.1438247 4.0609575 3.6610218 3.990
6912 3.7201951
377 378 379 380 381 382
383 384
3.9047793 4.0043710 3.7900550 3.7664826 3.7860664 3.9740014 4.112
6983 3.8775504
385 386 387 388 389 390
391 392
3.5993225 3.7980330 3.7137339 3.6672411 3.8328470 4.0043873 3.901
9512 3.8641798
393 394 395 396 397 398
399 400
3.9349736 3.6493487 3.8042936 3.8440669 3.7768278 4.0539993 4.169
2306 3.8499603
401 402 403 404 405 406
407 408
3.7939471 3.7575509 3.8140644 3.6013051 3.7786070 3.7960283 3.965
1654 3.5505769
409 410 411 412 413 414
415 416
3.8794206 3.9012051 3.8868614 4.1558967 3.7261709 3.7156750 3.736
8455 3.7200520
417 418 419 420 421 422
423 424
3.7291097 3.9154629 4.0297597 3.8560028 3.7052864 3.8037267 3.449
3253 3.7097445
425 426 427 428 429 430
431 432
4.0100855 3.7535633 3.8134097 3.8741604 3.9062042 3.8764783 3.682
5198 3.6802314
434 435 436 437 438 439
440 441
3.6883364 3.7630008 3.9687045 3.8211138 3.7806848 3.9010605 3.737
2589 3.9371863
442 443 444 445 446 447
448 449
3.8276090 0.9823738 4.0441106 3.9599217 3.8830140 3.7637120 4.125
0586 3.8974854
450 451 452 453 454 455
456 457
3.6093079 3.9360899 3.7791817 3.9838334 3.9150626 3.7105844 3.735
8467 3.8178523
458 459 460 461 462 463
464 465
3.9463559 3.8599338 3.7860598 3.9960874 3.8181361 3.8437134 3.749
5752 3.8224442

##	466	467	468	469	470	471
472	473					
##	3.7398341	3.9541891	3.8930094	3.8964411	3.7119239	3.7751380
4819	3.7966275					3.685
##	474	475	476	477	478	479
480	481					
##	4.1692306	3.8913453	4.1025907	3.7367166	3.8963265	3.6954433
8230	3.8942793					3.844
##	482	483	484	485	486	487
488	489					
##	3.8820210	4.1881064	3.8388052	3.7798145	3.7618541	3.7433108
0693	3.8582783					3.768
##	490	491	492	493	494	495
496	497					
##	4.1159924	3.6302101	3.8450604	3.9892586	4.1115490	3.9142533
0476	3.8641226					3.743
##	498	499	500	501	502	503
504	505					
##	3.8817317	3.9661300	3.5125971	3.6283829	3.7234760	4.0068274
7828	3.5539087					3.759
##	506	507	508	509	510	511
512	513					
##	3.8800269	3.8245496	3.5285808	3.9035738	4.1685736	3.9046660
1300	3.9354206					3.884
##	514	515	516	517	518	519
520	521					
##	3.8709438	4.0410369	3.7770101	3.8677410	3.8985631	4.0235624
0194	3.9254574					3.799
##	522	523	524	525	526	527
528	529					
##	4.0201876	3.9953602	3.8691783	4.0972952	3.8351303	4.0776752
7278	4.0284552					3.726
##	530	531	532	533	534	535
536	537					
##	3.9887789	3.6606500	3.9454999	4.0610324	3.9855277	3.7114761
6042	3.9074890					3.869
##	538	539	540	541	542	543
544	545					
##	3.8705495	3.8573809	4.2270409	3.8499471	3.7201203	3.9004886
5711	3.7695268					3.728
##	546	547	548	549	550	551
552	553					
##	3.9978856	4.0730913	4.0240549	3.8913453	3.9202595	3.8582687
3366	3.7105587					3.825
##	554	555	556	557	558	559
560	561					
##	3.8686009	3.8443541	4.0556555	3.7428088	3.8582783	3.8759934
1175	3.9652925					3.540
##	562	563	564	565	566	567

```

568      569
## 3.9724857 3.9016947 4.0344900 3.7999315 3.9435816 4.0171492 3.963
9673 3.6172819
##      570      571      572      573      574      575
576      577
## 3.7936578 3.8306070 3.8526718 4.2052817 3.9726327 4.1194947 4.080
1595 3.8791969
##      578      579      580      581      582      583
584      585
## 3.9908318 4.1224928 3.9239844 4.2308162 3.9326076 3.7816372 4.127
1743 3.7674923
##      586      587      588      589      590      591
592      593
## 4.0038596 3.9037124 3.9391498 4.0878567 3.7963207 3.8682454 3.899
6517 3.9516314
##      594      595      596      597      598      599
600      601
## 3.7982612 3.9315749 3.9199874 3.8918417 3.9304739 3.8912661 3.773
0379 3.6793382
##      602      603
## 3.7316017 3.9646598

```

residuals(fit)

```

##      1      2      3      4      5
6
## 1.254158731 1.443749024 1.066332694 0.415923913 0.258987616
0.272707106
##      7      8      9     10     11
12
## 0.072119785 0.266465362 0.040042660 0.055202947 0.254426819
0.264958539
##     13     14     15     16     17
18
## 0.176318345 0.243197265 0.540769662 0.270863006 -0.093089945
0.523552670
##     19     20     21     22     23
24
## 0.459435624 -0.159727490 0.235617285 0.137060678 0.053874975
0.209958320
##     25     26     27     28     29
30
## 0.376212804 0.331865007 0.291886605 0.257912948 0.204258291
0.075740896
##     31     32     33     34     35
36
## -0.091459019 0.171816623 0.002147606 0.198643027 0.361066570
-0.060784556
##     37     38     39     40     41
42

```


##	0.025128118	-0.715889393	-0.528858846	-0.679364639	-0.801069269
	-0.971372878				
##	43	44	45	46	47
48					
##	-0.758198483	-0.866774363	-0.853456110	-0.890783175	-0.674283204
	-0.841288380				
##	49	50	51	52	53
54					
##	-0.369482396	-1.768855359	-2.807971830	1.145251587	1.101787558
	0.301053475				
##	55	56	57	58	59
60					
##	0.258987616	0.170432550	0.266941093	0.043901332	0.285705904
	0.079295636				
##	61	62	63	64	65
66					
##	0.164244814	0.299189506	0.352377232	0.501286735	0.264958539
	0.115766731				
##	67	68	69	70	71
72					
##	0.143216652	0.170142676	0.174451311	0.379573416	0.301687610
	0.298437025				
##	73	74	75	76	77
78					
##	0.179872255	0.193434773	-0.053778523	0.175541065	0.467386498
	0.312423066				
##	79	80	81	82	83
84					
##	0.197560586	0.044200926	0.053223683	0.338494545	0.332588052
	0.255827186				
##	85	86	87	88	89
90					
##	0.165911745	0.079013979	0.328446542	0.185723431	-0.863284690
	-0.884605524				
##	91	92	93	94	95
96					
##	-0.777876556	-0.596738219	-0.873408076	-0.674283204	-0.980143834
	-0.369482396				
##	97	98	99	100	101
102					
##	-0.816692429	-1.794842041	-1.858282145	-1.836262360	-1.560337767
	-1.815240324				
##	103	104	105	106	107
108					
##	-1.733454694	-2.849347709	1.206587577	0.046815201	0.293516668
	0.158565131				
##	109	110	111	112	113
114					
##	-0.039970836	0.042335493	0.012370570	-0.093937258	0.094597174

0.132563821					
##	115	116	117	118	119
120					
##	0.334184449	0.221919426	-0.092721356	0.104585179	0.232056254
0.124033681					
##	121	122	123	124	125
126					
##	0.280687548	0.161598342	0.387805076	-0.097710379	0.311458510
-0.049851254					
##	127	128	129	130	131
132					
##	0.340383322	0.120390745	0.152527225	0.260240759	0.350399624
0.248898731					
##	133	134	135	136	137
138					
##	0.242681990	0.175330996	-0.731515792	-0.907588426	-0.896693357
-1.533286547					
##	139	140	141	142	143
144					
##	-2.687271695	1.271347334	1.350771727	1.200709732	1.108350706
1.292227753					
##	145	146	147	148	149
150					
##	0.184462520	0.149563010	0.124323670	0.484263516	-0.057993289
0.152953770					
##	151	152	153	154	155
156					
##	0.042730203	0.207278092	0.132563821	0.076433316	0.214150791
0.282629276					
##	157	158	159	160	161
162					
##	0.174806019	0.228032191	-0.073804936	0.127973377	0.180066333
0.301795030					
##	163	164	165	166	167
168					
##	0.491434689	0.437587198	0.195394070	0.366030925	0.126632880
0.151908536					
##	169	170	171	172	173
174					
##	0.270578763	0.093633782	0.247952194	0.232974279	0.146945379
0.147334001					
##	175	176	177	178	179
180					
##	0.049348641	0.110920481	0.080927699	0.260363878	0.103101442
0.099549240					
##	181	182	183	184	185
186					
##	0.097878294	0.258499670	0.194115239	0.250667279	0.166149075
0.204697073					

##	187	188	189	190	191
192					
##	-0.091073131	-0.539044293	-0.337863191	-0.773259808	-0.812223547
	-0.686439789				
##	193	194	195	196	197
198					
##	-0.798427891	-0.714877179	-0.613940487	-0.898529110	-0.730184095
	-0.693083904				
##	199	200	201	202	203
204					
##	-0.847170020	-0.868439351	-0.684334333	-0.910658590	-0.952535009
	-0.701664028				
##	205	206	207	208	209
210					
##	-0.841889857	-0.828528733	-0.803402246	-1.790484257	-1.835944105
	-1.683186064				
##	211	212	213	214	215
216					
##	1.000957241	0.902444014	1.234345046	0.266441994	0.137400684
	0.210144999				
##	217	218	219	220	221
222					
##	0.039890581	0.294214253	0.095727324	-0.047321225	0.222288884
	0.185321481				
##	223	224	225	226	227
228					
##	0.166297495	0.058390095	0.380147873	0.279645893	0.211337736
	0.210328333				
##	229	230	231	232	233
234					
##	0.268055201	0.298077973	-0.047836180	0.091177781	0.102950515
	0.080784434				
##	235	236	237	238	239
240					
##	0.219660540	0.127185145	0.278752355	0.087004441	0.283616594
	0.103101442				
##	241	242	243	244	245
246					
##	0.102641581	0.538476959	0.056671429	0.254175178	0.036971810
	0.175411220				
##	247	248	249	250	251
252					
##	0.264765099	0.122647498	0.107080658	-0.105523201	0.271170330
	-0.842449347				
##	253	254	255	256	257
258					
##	-0.839452815	-0.519374088	-0.689172621	-1.101699490	-0.625870560
	-0.940784990				
##	259	260	261	262	263

264
-0.934524537 -0.866257267 -1.091293773 -0.918449549 -1.878600819
-1.988611544
265 266 267 268 269
270
-1.850243376 -1.768362927 -1.792446218 -2.652862886 1.047430600
1.029436209
271 272 273 274 275
276
1.156836146 1.212283135 1.192180172 1.157679545 1.328468977
1.081069014
277 278 279 280 281
282
0.670458672 0.305828444 -0.075766661 0.091153115 0.209945040
0.088999135
283 284 285 286 287
288
0.060549288 0.070513250 0.141809085 0.169442442 0.186876180
0.159431982
289 290 291 292 293
294
0.318809895 0.272566091 0.069822676 0.072455793 0.112737764
0.352536735
295 296 297 298 299
300
0.090072465 0.118568523 0.170687447 0.071976253 0.251615505
0.078710129
301 302 303 304 305
306
0.328738586 0.223172187 0.015434643 0.071135265 0.038499691
0.068492023
307 308 309 310 311
312
0.372180564 0.143074596 0.259812700 0.173984696 0.398694944
-0.037809317
313 314 315 316 317
318
0.196704860 0.119813868 0.149386293 0.221393014 0.308193486
0.377701649
319 320 321 322 323
324
0.308369055 0.364562298 0.199494534 0.143265763 0.328468977
0.371126105
325 326 327 328 329
330
0.114930657 0.202623324 -0.050451642 0.232125042 0.247795608
0.086322328
331 332 333 334 335
336

##	-0.143625483	0.320564986	0.552426674	0.035620811	0.141099145
	0.269793703				
##	337	338	339	340	341
	342				
##	0.234877354	0.191828996	-0.799265883	-0.889586434	-0.882279914
	-0.739989991				
##	343	344	345	346	347
	348				
##	-0.735868419	-0.805091348	-0.806033956	-0.673929080	-0.809973735
	-1.086649349				
##	349	350	351	352	353
	354				
##	-0.876886834	-0.666613309	-0.918148416	-0.702942901	-0.639223281
	-0.807114942				
##	355	356	357	358	359
	360				
##	-0.727593777	-0.842755740	-0.835607908	-0.735468656	-1.964742577
	-1.851484841				
##	361	362	363	364	365
	366				
##	-1.713053505	-3.061965083	-3.068436652	1.156836146	1.212283135
	1.133490046				
##	367	368	369	370	371
	372				
##	1.285558417	0.976291686	1.270480472	1.231012908	0.970614950
	0.856175299				
##	373	374	375	376	377
	378				
##	0.939042546	0.338978195	0.009308781	0.279804874	0.095220655
	-0.004370994				
##	379	380	381	382	383
	384				
##	0.209945040	0.233517395	0.213933557	0.025998588	-0.112698342
	0.122449597				
##	385	386	387	388	389
	390				
##	0.400677505	0.201966994	0.286266113	0.332758905	0.167152996
	-0.004387276				
##	391	392	393	394	395
	396				
##	0.098048795	0.135820225	0.065026443	0.350651340	0.195706433
	0.155933129				
##	397	398	399	400	401
	402				
##	0.223172187	-0.053999277	-0.169230593	0.150039706	0.206052880
	0.242449065				
##	403	404	405	406	407
	408				
##	0.185935567	0.398694944	0.221393014	0.203971725	0.034834552

0.449423149					
##	409	410	411	412	413
414					
##	0.120579436	0.098794878	0.113138636	-0.155896743	0.273829120
0.284325045					
##	415	416	417	418	419
420					
##	0.263154470	0.279947980	0.270890264	0.084537090	-1.029759743
-0.856002834					
##	421	422	423	424	425
426					
##	-0.705286364	-0.803726720	-0.449325322	-0.709744475	-1.010085539
-0.753563336					
##	427	428	429	430	431
432					
##	-0.813409679	-0.874160448	-0.906204244	-0.876478324	-0.682519753
-0.680231396					
##	434	435	436	437	438
439					
##	-0.688336375	-0.763000762	-0.968704460	-1.821113801	-1.780684788
-1.901060479					
##	440	441	442	443	444
445					
##	-1.737258932	-1.937186348	-2.827609013	0.017626234	0.955889355
1.040078336					
##	446	447	448	449	450
451					
##	1.116986016	1.236287968	0.874941379	1.102514576	1.390692113
1.063910129					
##	452	453	454	455	456
457					
##	1.220818310	1.016166574	0.084937365	0.289415561	0.264153278
0.182147668					
##	458	459	460	461	462
463					
##	0.053644082	0.140066208	0.213940188	0.003912639	0.181863869
0.156286595					
##	464	465	466	467	468
469					
##	0.250424840	0.177555751	0.260165886	0.045810898	0.106990575
0.103558856					
##	470	471	472	473	474
475					
##	0.288076096	0.224862039	0.314518088	0.203372488	-0.169230593
0.108654697					
##	476	477	478	479	480
481					
##	-0.102590731	0.263283428	0.103673476	0.304556689	0.155177009
0.105720742					

##	482	483	484	485	486
487					
##	0.117979005	-0.188106412	0.161194828	0.220185485	0.238145883
0.256689166					
##	488	489	490	491	492
493					
##	0.231930663	0.141721698	-0.115992368	0.369789911	0.154939586
-0.989258627					
##	494	495	496	497	498
499					
##	-1.111549041	-0.914253273	-0.743047600	-0.864122607	-0.881731716
-0.966129989					
##	500	501	502	503	504
505					
##	-0.512597141	-0.628382869	-0.723476037	-1.006827388	-0.759782768
-0.553908717					
##	506	507	508	509	510
511					
##	-0.880026852	-0.824549583	-0.528580805	1.096426224	0.831426397
1.095333991					
##	512	513	514	515	516
517					
##	1.115869966	1.064579412	1.129056203	0.958963086	1.222989900
1.132259012					
##	518	519	520	521	522
523					
##	1.101436895	0.976437614	1.200980635	1.074542574	0.979812359
1.004639766					
##	524	525	526	527	528
529					
##	1.130821725	0.902704796	1.164869676	-0.077675199	0.273272198
-0.028455167					
##	530	531	532	533	534
535					
##	0.011221069	0.339350010	0.054500120	-0.061032442	0.014472266
0.288523881					
##	536	537	538	539	540
541					
##	0.130395782	0.092510974	0.129450516	0.142619120	-0.227040895
0.150052885					
##	542	543	544	545	546
547					
##	0.279879681	0.099511367	0.271428918	0.230473206	0.002114440
-0.073091300					
##	548	549	550	551	552
553					
##	-0.024054921	0.108654697	0.079740482	0.141731339	0.174663378
0.289441349					
##	554	555	556	557	558

```

559
## 0.131399114 0.155645888 -0.055655522 0.257191213 0.141721698
0.124006628
## 560 561 562 563 564
565
## 0.459882532 0.034707467 0.027514332 0.098305312 -0.034490039
0.200068519
## 566 567 568 569 570
571
## -0.943581609 -1.017149236 -0.963967251 -0.617281922 -0.793657756
-0.830606990
## 572 573 574 575 576
577
## -0.852671763 0.794718281 1.027367312 0.880505272 0.919840547
1.120803114
## 578 579 580 581 582
583
## 1.009168230 0.877507167 1.076015615 0.769183770 1.067392432
1.218362814
## 584 585 586 587 588
589
## 0.872825660 1.232507726 0.996140441 1.096287600 1.060850233
0.912143259
## 590 591 592 593 594
595
## 1.203679290 1.131754642 1.100348310 1.048368584 1.201738819
1.068425056
## 596 597 598 599 600
601
## 0.080012581 0.108158270 0.069526080 0.108733873 0.226962099
0.320661802
## 602 603
## 0.268398270 0.035340170

```

```
#Anova Table
```

```
anova(fit)
```

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

```
##
```

```
## Response: rating
```

```

##      Df Sum Sq Mean Sq F value    Pr(>F)
## bpm      1    0.28   0.2844   0.5130   0.47413
## nrgy      1    1.82   1.8160   3.2756   0.07082 .
## dnce      1    4.17   4.1702   7.5218   0.00628 **
## dB        1   11.34  11.3374  20.4495 7.396e-06 ***
## val       1    0.01   0.0142   0.0256   0.87293
## dur       1    2.94   2.9404   5.3036   0.02163 *
## acous      1    0.02   0.0210   0.0379   0.84562
## spch       1    0.03   0.0282   0.0508   0.82173
## Residuals 593 328.77   0.5544

```



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

vcov(fit)

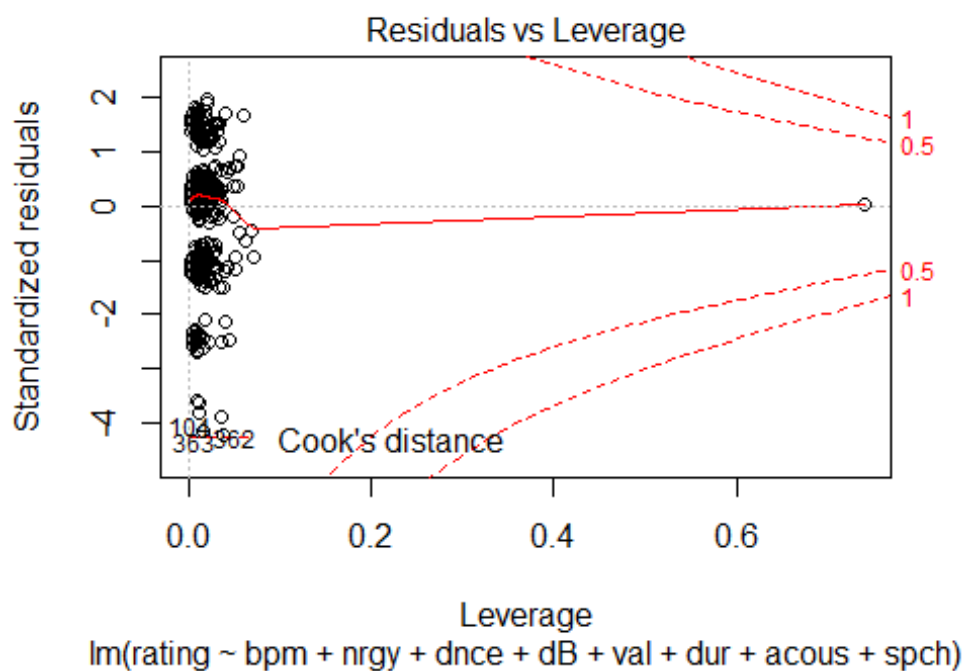
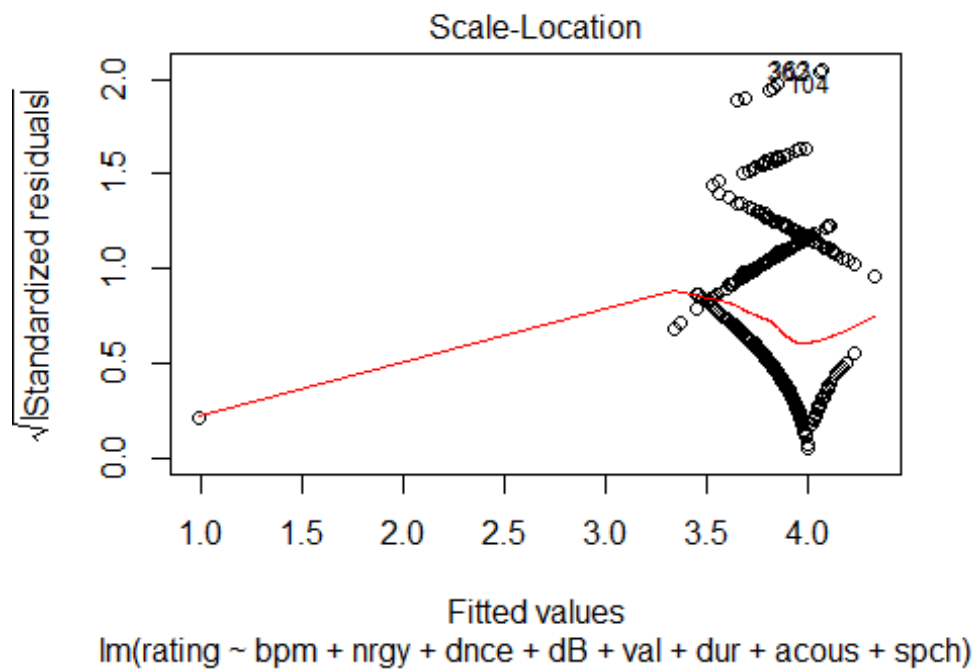
##              (Intercept)              bpm              nrgy              dn
ce
## (Intercept)  2.070199e-01 -2.796029e-04 -7.902573e-04 -6.939011e-
04
## bpm          -2.796029e-04  1.643313e-06  1.875027e-07  7.121814e-
07
## nrgy         -7.902573e-04  1.875027e-07  8.024808e-06  1.721499e-
06
## dn          -6.939011e-04  7.121814e-07  1.721499e-06  7.853443e-
06
## dB           3.404769e-03 -3.422102e-06 -1.979914e-05 -8.137665e-
06
## val          4.754108e-05 -7.671228e-08 -1.434999e-06 -2.276277e-
06
## dur          -2.287651e-04  3.983176e-08  1.130329e-07  1.475965e-
07
## acous       -3.905811e-04  2.999826e-07  2.927088e-06  1.230435e-
06
## spch         1.095720e-04 -3.185587e-07 -1.350512e-06  6.577942e-
07
##              dB              val              dur              aco
us
## (Intercept)  3.404769e-03  4.754108e-05 -2.287651e-04 -3.905811e-
04
## bpm          -3.422102e-06 -7.671228e-08  3.983176e-08  2.999826e-
07
## nrgy         -1.979914e-05 -1.434999e-06  1.130329e-07  2.927088e-
06
## dn          -8.137665e-06 -2.276277e-06  1.475965e-07  1.230435e-
06
## dB           1.856477e-04  3.019156e-07 -8.925313e-08 -5.609375e-
06
## val          3.019156e-07  3.021001e-06  2.852947e-07 -2.179137e-
07
## dur          -8.925313e-08  2.852947e-07  8.623461e-07  2.393589e-
08
## acous       -5.609375e-06 -2.179137e-07  2.393589e-08  3.454919e-
06
## spch         4.936074e-06 -9.525799e-07 -3.433234e-07 -6.311856e-
07
##              spch
## (Intercept)  1.095720e-04
## bpm          -3.185587e-07
## nrgy         -1.350512e-06
## dn          6.577942e-07
```

```
## dB          4.936074e-06
## val         -9.525799e-07
## dur         -3.433234e-07
## acous       -6.311856e-07
## spch        1.730875e-05
```

```
library(car)
```

```
#diagnostic plots
```

```
plot(fit)
```

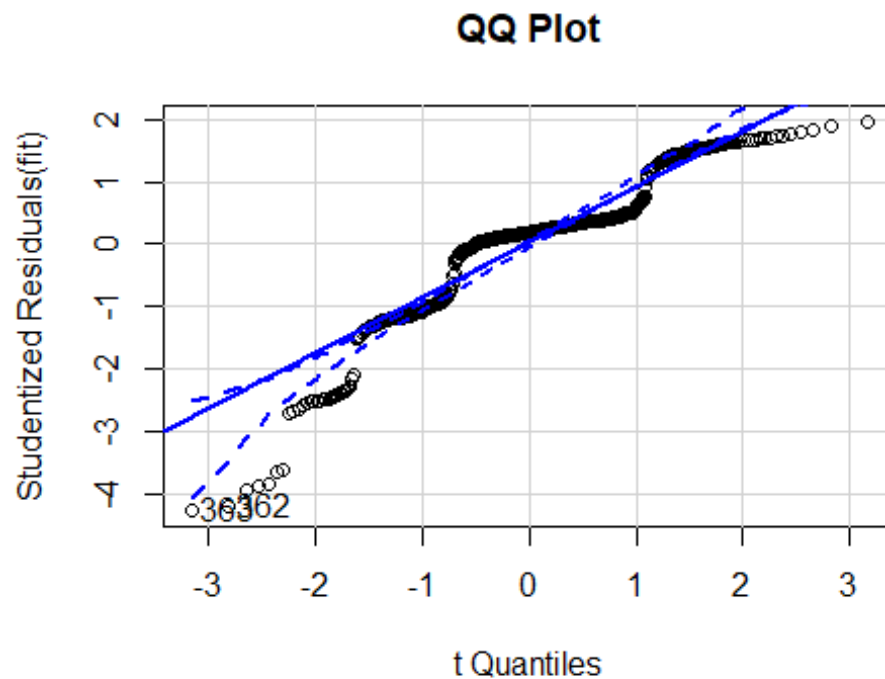



Assessing Outliers

`outlierTest(fit)`

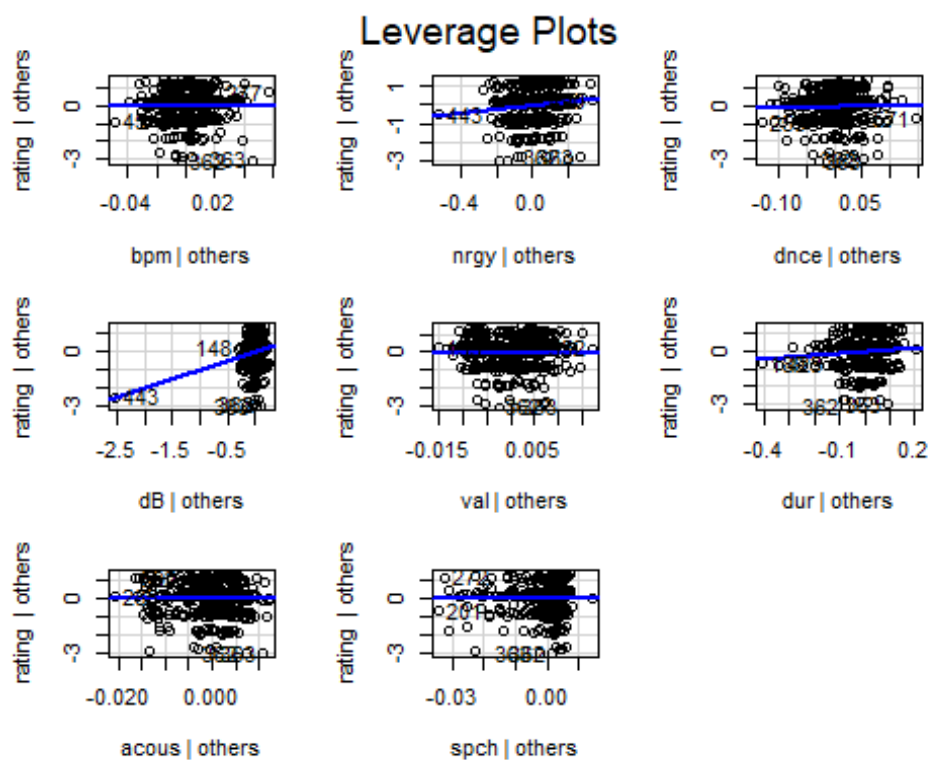
```
##      rstudent unadjusted p-value Bonferroni p
## 363 -4.265552      2.3211e-05      0.013973
## 362 -4.205441      3.0092e-05      0.018115
```

```
qqPlot(fit, main="QQ Plot")
```



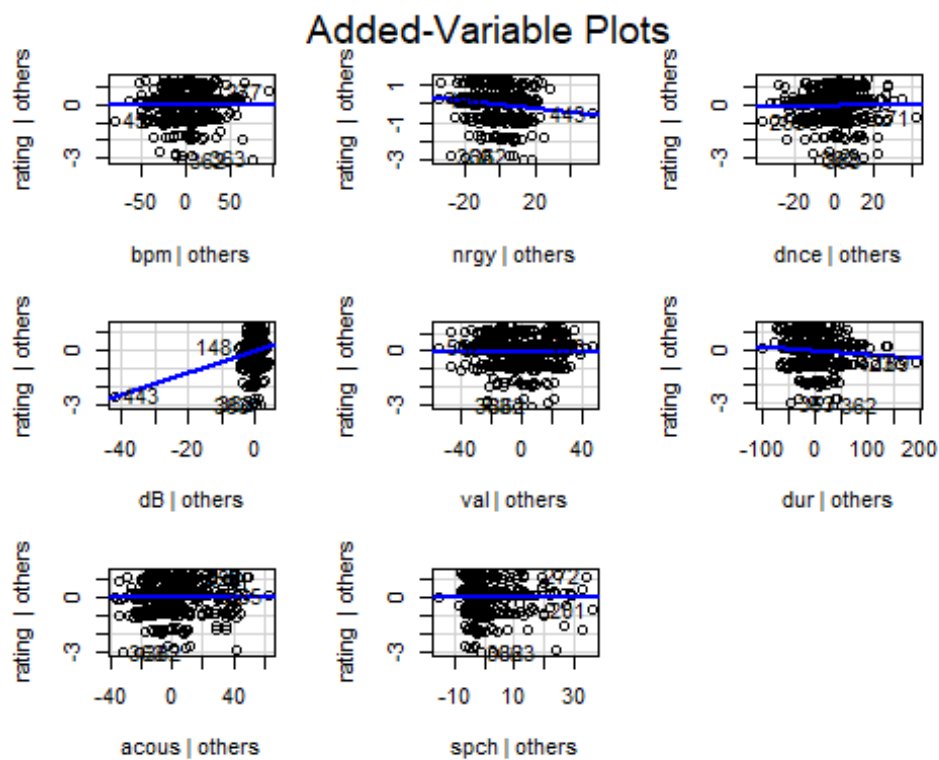
```
## [1] 362 363
```

```
leveragePlots(fit) # Leverage plots
```

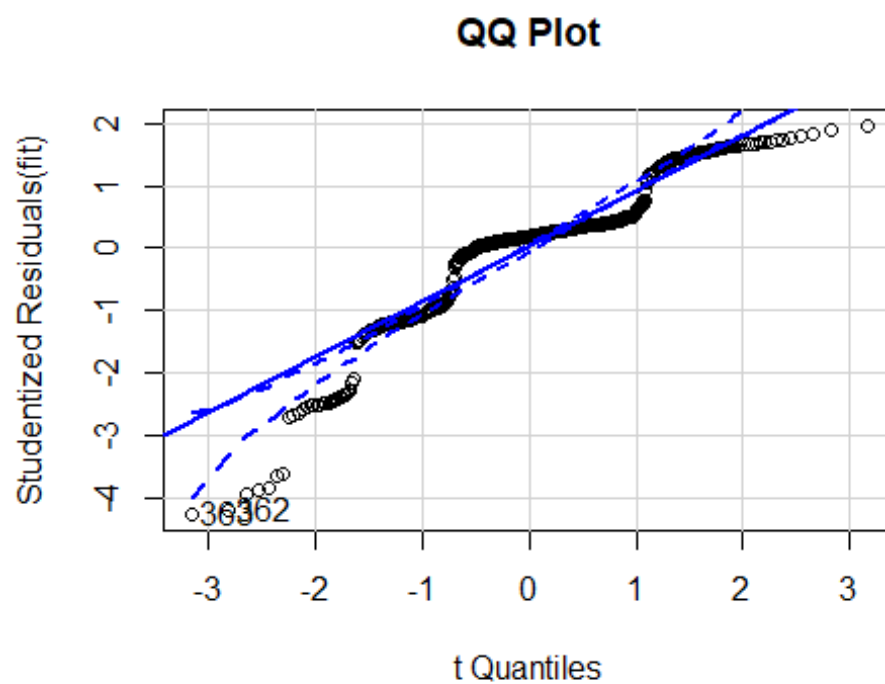


```
# Influential Observations
```

```
# added variable plots
avPlots(fit)
```

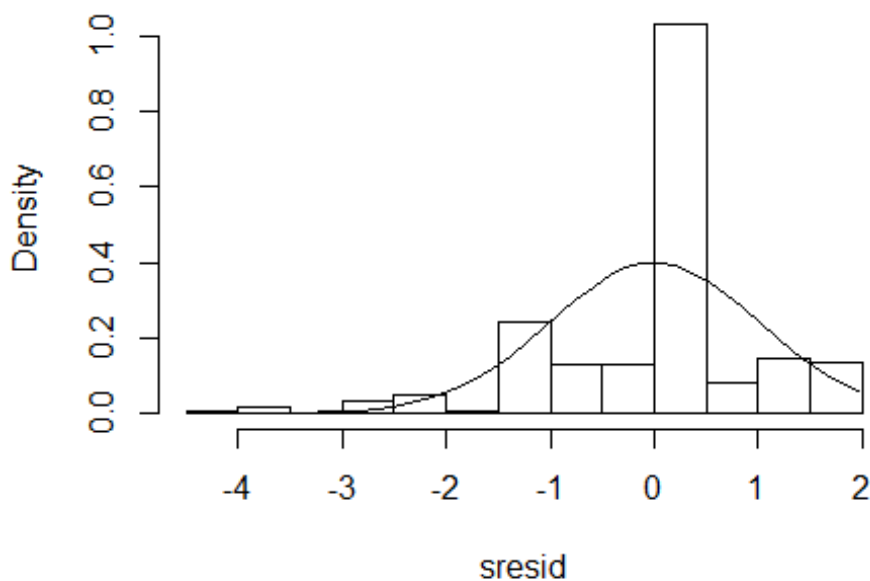


```
# Normality of Residuals
# qq plot for studentized resid
qqPlot(fit, main="QQ Plot")
```



```
## [1] 362 363
# distribution of studentized residuals
library(MASS)
sresid <- studres(fit)
hist(sresid, freq=FALSE,
     main="Distribution of Studentized Residuals")
xfit<-seq(min(sresid),max(sresid),length=40)
yfit<-dnorm(xfit)
lines(xfit, yfit)
```

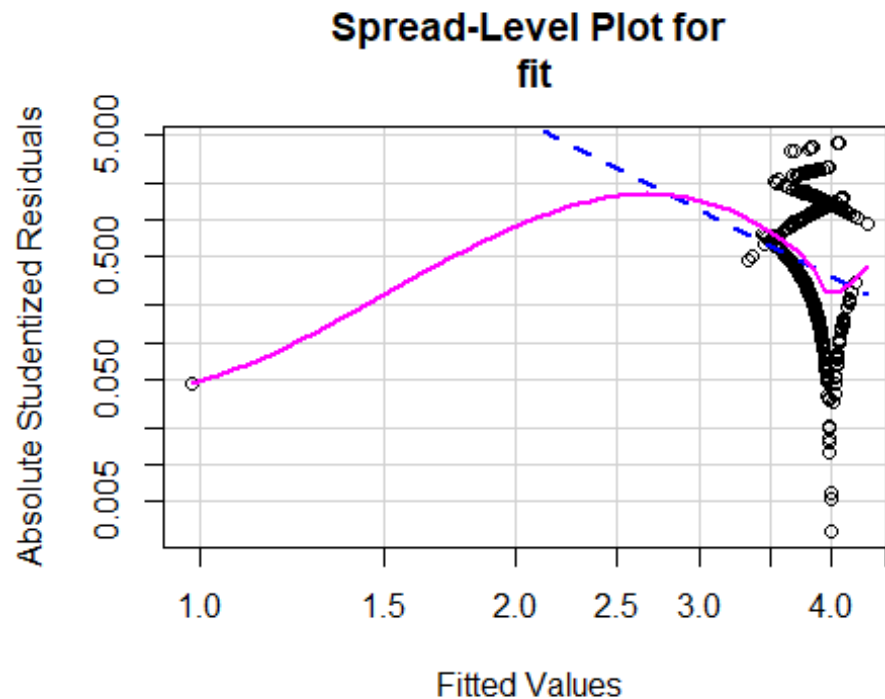
Distribution of Studentized Residuals



```
#Non-constant Error Variance
# Evaluate homoscedasticity
# non-constant error variance test
ncvTest(fit)

## Non-constant Variance Score Test
## Variance formula: ~ fitted.values
## Chisquare = 1.648805, Df = 1, p = 0.19912

# plot studentized residuals vs. fitted values
spreadLevelPlot(fit)
```



```
##
## Suggested power transformation: 5.341688

#Multi-collinearity
# Evaluate Collinearity
vif(fit) # variance inflation factors

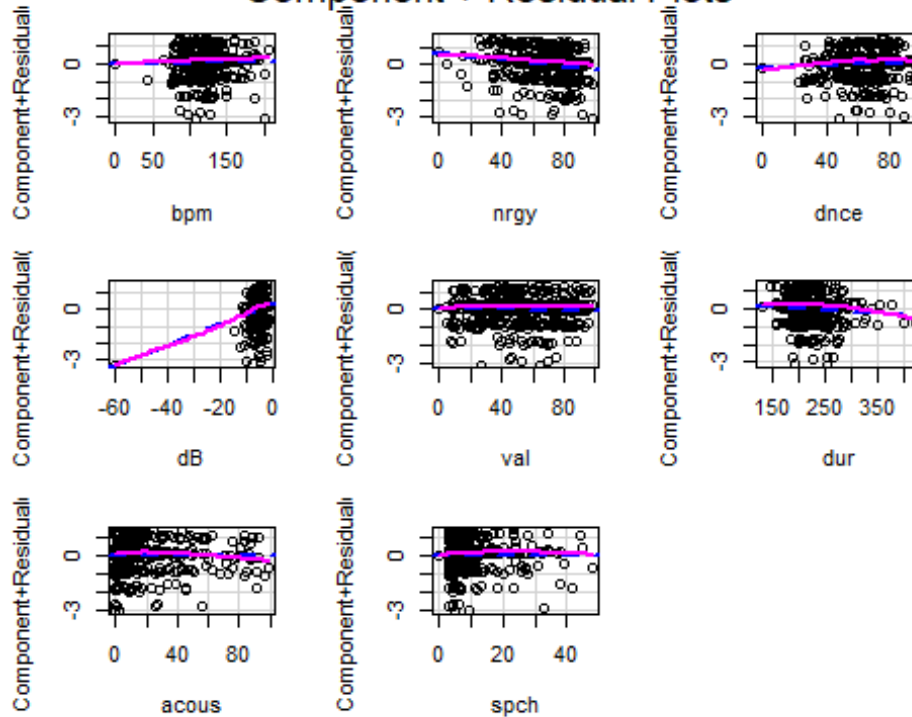
##      bpm      nrgy      dnce      dB      val      dur      acous
spch
## 1.094630 2.317866 1.521431 1.577344 1.660351 1.088451 1.617399 1.
052035

sqrt(vif(fit)) > 2 # problem? ##tells disp and wt are correlated a
nd one can be dropped

##      bpm      nrgy      dnce      dB      val      dur      acous      spch
## FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE

#Nonlinearity
# component + residual plot
crPlots(fit)
```


Component + Residual Plots



```
library(gvlma)
gvmodel <- gvlma(fit)
summary(gvmodel)

##
## Call:
## lm(formula = rating ~ bpm + nrgy + dnce + dB + val + dur + acous +
##     spch, data = data_clean)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.0684 -0.1588  0.1418  0.2854  1.4438
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  5.0729840   0.4549944   11.150  < 2e-16 ***
## bpm           0.0006058   0.0012819    0.473  0.636706
## nrgy          -0.0099612   0.0028328   -3.516  0.000471 ***
## dnce           0.0035088   0.0028024    1.252  0.211032
## dB             0.0601839   0.0136253    4.417  1.19e-05 ***
## val           -0.0003163   0.0017381   -0.182  0.855654
## dur           -0.0021127   0.0009286   -2.275  0.023259 *
## acous         -0.0003267   0.0018587   -0.176  0.860553
## spch          -0.0009379   0.0041604   -0.225  0.821725
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
```

```

## Residual standard error: 0.7446 on 593 degrees of freedom
## Multiple R-squared:  0.059, Adjusted R-squared:  0.0463
## F-statistic: 4.647 on 8 and 593 DF,  p-value: 1.587e-05
##
##
## ASSESSMENT OF THE LINEAR MODEL ASSUMPTIONS
## USING THE GLOBAL TEST ON 4 DEGREES-OF-FREEDOM:
## Level of Significance =  0.05
##
## Call:
## gvlma(x = fit)
##
##
## Value p-value Decision
## Global Stat      216.11503  0.0000 Assumptions NOT satisfied!
## Skewness         105.55254  0.0000 Assumptions NOT satisfied!
## Kurtosis         109.48132  0.0000 Assumptions NOT satisfied!
## Link Function      0.04449  0.8329 Assumptions acceptable.
## Heteroscedasticity 1.03668  0.3086 Assumptions acceptable.

fit

##
## Call:
## lm(formula = rating ~ bpm + nrgy + dnce + dB + val + dur + acous
+
##     spch, data = data_clean)
##
## Coefficients:
## (Intercept)          bpm          nrgy          dnce          dB
val
##  5.0729840    0.0006058   -0.0099612    0.0035088    0.0601839
-0.0003163
##          dur          acous          spch
## -0.0021127   -0.0003267   -0.0009379

summary(fit)

##
## Call:
## lm(formula = rating ~ bpm + nrgy + dnce + dB + val + dur + acous
+
##     spch, data = data_clean)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.0684 -0.1588  0.1418  0.2854  1.4438
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  5.0729840   0.4549944   11.150  < 2e-16 ***

```

```

## bpm          0.0006058  0.0012819   0.473 0.636706
## nrgy         -0.0099612  0.0028328  -3.516 0.000471 ***
## dnce          0.0035088  0.0028024   1.252 0.211032
## dB           0.0601839  0.0136253   4.417 1.19e-05 ***
## val          -0.0003163  0.0017381  -0.182 0.855654
## dur          -0.0021127  0.0009286  -2.275 0.023259 *
## acous        -0.0003267  0.0018587  -0.176 0.860553
## spch         -0.0009379  0.0041604  -0.225 0.821725
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.7446 on 593 degrees of freedom
## Multiple R-squared:  0.059, Adjusted R-squared:  0.0463
## F-statistic: 4.647 on 8 and 593 DF, p-value: 1.587e-05

fit1 <- fit
fit2 <- lm(rating~bpm+nrgy+dnce+dB+val+dur+spch,data = data_clean)
summary(fit2)

##
## Call:
## lm(formula = rating ~ bpm + nrgy + dnce + dB + val + dur + spch,
##     data = data_clean)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.0579 -0.1595  0.1426  0.2875  1.4353
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  5.0360542  0.4032353  12.489 < 2e-16 ***
## bpm          0.0006341  0.0012707   0.499  0.6179
## nrgy        -0.0096844  0.0023528  -4.116 4.40e-05 ***
## dnce         0.0036252  0.0027209   1.332  0.1833
## dB          0.0596535  0.0132760   4.493 8.43e-06 ***
## val        -0.0003369  0.0017327  -0.194  0.8459
## dur        -0.0021104  0.0009278  -2.275  0.0233 *
## spch        -0.0009975  0.0041431  -0.241  0.8098
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.744 on 594 degrees of freedom
## Multiple R-squared:  0.05895, Adjusted R-squared:  0.04786
## F-statistic: 5.315 on 7 and 594 DF, p-value: 6.532e-06

fit3 <- lm(rating~bpm+nrgy+dnce+dB+dur+spch,data = data_clean)
summary(fit3)

##
## Call:

```

```
## lm(formula = rating ~ bpm + nrgy + dnce + dB + dur + spch, data =
data_clean)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.0567 -0.1639  0.1426  0.2873  1.4374
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  5.0386204  0.4026932  12.512  < 2e-16 ***
## bpm          0.0006277  0.0012692   0.495  0.6211
## nrgy        -0.0098245  0.0022380  -4.390 1.34e-05 ***
## dnce         0.0033789  0.0024060   1.404  0.1607
## dB           0.0596477  0.0132652   4.497 8.31e-06 ***
## dur         -0.0020783  0.0009122  -2.278  0.0231 *
## spch        -0.0011087  0.0041001  -0.270  0.7869
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.7434 on 595 degrees of freedom
## Multiple R-squared:  0.05889,    Adjusted R-squared:  0.0494
## F-statistic: 6.205 on 6 and 595 DF,  p-value: 2.484e-06

fit4 <- lm(rating~bpm+nrgy+dnce+dB+dur,data = data_clean)
summary(fit4)

##
## Call:
## lm(formula = rating ~ bpm + nrgy + dnce + dB + dur, data = data_c
lean)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.0576 -0.1620  0.1412  0.2872  1.4227
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  5.0416295  0.4022263  12.534  < 2e-16 ***
## bpm          0.0006091  0.0012664   0.481  0.6307
## nrgy        -0.0099053  0.0022163  -4.469 9.39e-06 ***
## dnce         0.0033892  0.0024038   1.410  0.1591
## dB           0.0599037  0.0132211   4.531 7.10e-06 ***
## dur         -0.0020943  0.0009096  -2.303  0.0216 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.7428 on 596 degrees of freedom
## Multiple R-squared:  0.05877,    Adjusted R-squared:  0.05088
## F-statistic: 7.443 on 5 and 596 DF,  p-value: 8.704e-07
```

```

fit5 <- lm(rating~nrgy+dnce+dB+dur,data = data_clean)
summary(fit5)

##
## Call:
## lm(formula = rating ~ nrgy + dnce + dB + dur, data = data_clean)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.0543 -0.1643  0.1431  0.2912  1.4047
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  5.1340372  0.3531439  14.538  < 2e-16 ***
## nrgy        -0.0098632  0.0022131  -4.457  9.94e-06 ***
## dnce         0.0031754  0.0023608   1.345   0.1791
## dB           0.0609888  0.0130188   4.685  3.48e-06 ***
## dur         -0.0021091  0.0009085  -2.322   0.0206 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.7423 on 597 degrees of freedom
## Multiple R-squared:  0.05841,    Adjusted R-squared:  0.0521
## F-statistic: 9.258 on 4 and 597 DF,  p-value: 2.91e-07

fit6 <- lm(rating~nrgy+dB+dur,data = data_clean)
summary(fit6)

##
## Call:
## lm(formula = rating ~ nrgy + dB + dur, data = data_clean)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.0543 -0.1450  0.1479  0.2858  1.4418
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  5.3903496  0.2975144  18.118  < 2e-16 ***
## nrgy        -0.0097587  0.0022133  -4.409  1.23e-05 ***
## dB           0.0639368  0.0128417   4.979  8.38e-07 ***
## dur         -0.0023001  0.0008979  -2.562   0.0107 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.7428 on 598 degrees of freedom
## Multiple R-squared:  0.05555,    Adjusted R-squared:  0.05081
## F-statistic: 11.72 on 3 and 598 DF,  p-value: 1.791e-07

```

```
# compare models
```

```
anova(fit1, fit6)
```

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

```
##
```

```
## Model 1: rating ~ bpm + nrgy + dnce + dB + val + dur + acous + sp  
ch
```

```
## Model 2: rating ~ nrgy + dB + dur
```

```
##   Res.Df    RSS Df Sum of Sq    F Pr(>F)
```

```
## 1      593 328.77
```

```
## 2      598 329.97 -5    -1.2029 0.434  0.825
```

```
step <- stepAIC(fit1, direction="both")
```

```
## Start:  AIC=-346.16
```

```
## rating ~ bpm + nrgy + dnce + dB + val + dur + acous + spch
```

```
##
```

```
##           Df Sum of Sq    RSS    AIC
```

```
## - acous  1      0.0171 328.78 -348.13
```

```
## - val    1      0.0184 328.78 -348.12
```

```
## - spch   1      0.0282 328.79 -348.11
```

```
## - bpm    1      0.1238 328.89 -347.93
```

```
## - dnce   1      0.8692 329.63 -346.57
```

```
## <none>                328.77 -346.16
```

```
## - dur    1      2.8696 331.63 -342.93
```

```
## - nrgy   1      6.8552 335.62 -335.73
```

```
## - dB     1     10.8169 339.58 -328.67
```

```
##
```

```
## Step:  AIC=-348.13
```

```
## rating ~ bpm + nrgy + dnce + dB + val + dur + spch
```

```
##
```

```
##           Df Sum of Sq    RSS    AIC
```

```
## - val    1      0.0209 328.80 -350.09
```

```
## - spch   1      0.0321 328.81 -350.07
```

```
## - bpm    1      0.1379 328.92 -349.87
```

```
## - dnce   1      0.9826 329.76 -348.33
```

```
## <none>                328.78 -348.13
```

```
## + acous  1      0.0171 328.77 -346.16
```

```
## - dur    1      2.8640 331.65 -344.91
```

```
## - nrgy   1      9.3775 338.16 -333.20
```

```
## - dB     1     11.1753 339.96 -330.00
```

```
##
```

```
## Step:  AIC=-350.09
```

```
## rating ~ bpm + nrgy + dnce + dB + dur + spch
```

```
##
```

```
##           Df Sum of Sq    RSS    AIC
```

```
## - spch   1      0.0404 328.84 -352.01
```

```
## - bpm    1      0.1351 328.94 -351.84
```

```
## - dnce   1      1.0898 329.89 -350.10
```

```

## <none>          328.80 -350.09
## + val      1      0.0209 328.78 -348.13
## + acous    1      0.0197 328.78 -348.12
## - dur      1      2.8684 331.67 -346.86
## - nrgy     1     10.6492 339.45 -332.90
## - dB       1     11.1732 339.98 -331.97
##
## Step: AIC=-352.01
## rating ~ bpm + nrgy + dnce + dB + dur
##
##           Df Sum of Sq    RSS      AIC
## - bpm      1      0.1276 328.97 -353.78
## <none>          328.84 -352.01
## - dnce     1      1.0968 329.94 -352.01
## + spch     1      0.0404 328.80 -350.09
## + val      1      0.0292 328.81 -350.07
## + acous    1      0.0250 328.82 -350.06
## - dur      1      2.9253 331.77 -348.68
## - nrgy     1     11.0212 339.86 -334.17
## - dB       1     11.3270 340.17 -333.63
##
## Step: AIC=-353.78
## rating ~ nrgy + dnce + dB + dur
##
##           Df Sum of Sq    RSS      AIC
## - dnce     1      0.9969 329.97 -353.96
## <none>          328.97 -353.78
## + bpm      1      0.1276 328.84 -352.01
## + acous    1      0.0398 328.93 -351.85
## + spch     1      0.0329 328.94 -351.84
## + val      1      0.0253 328.95 -351.83
## - dur      1      2.9702 331.94 -350.37
## - nrgy     1     10.9448 339.92 -336.08
## - dB       1     12.0932 341.06 -334.05
##
## Step: AIC=-353.96
## rating ~ nrgy + dB + dur
##
##           Df Sum of Sq    RSS      AIC
## <none>          329.97 -353.96
## + dnce     1      0.9969 328.97 -353.78
## + acous    1      0.1669 329.80 -352.26
## + val      1      0.1028 329.87 -352.15
## + spch     1      0.0431 329.93 -352.04
## + bpm      1      0.0277 329.94 -352.01
## - dur      1      3.6210 333.59 -349.39
## - nrgy     1     10.7273 340.70 -336.70
## - dB       1     13.6782 343.65 -331.51

```

```
step$anova # display results
```

```
## Stepwise Model Path
```

```
## Analysis of Deviance Table
```

```
##
```

```
## Initial Model:
```

```
## rating ~ bpm + nrgy + dnce + dB + val + dur + acous + spch
```

```
##
```

```
## Final Model:
```

```
## rating ~ nrgy + dB + dur
```

```
##
```

```
##
```

##		Step	Df	Deviance	Resid. Df	Resid. Dev	AIC
##	1				593	328.7652	-346.1579
##	2	- acous	1	0.01712388	594	328.7824	-348.1266
##	3	- val	1	0.02092719	595	328.8033	-350.0882
##	4	- spch	1	0.04040772	596	328.8437	-352.0143
##	5	- bpm	1	0.12763240	597	328.9713	-353.7807
##	6	- dnce	1	0.99685615	598	329.9682	-353.9592

```
attach(data_clean)
```

```
predict.lm(fit6,data.frame(nrgy=86, dB = -3,dur=120))
```

```
## 1
```

```
## 4.083278
```

#As we can observe that the multiple regression model gives best value for f-stat when we take

independent variables as energy, dB as in Loudness and duration, the p-value obtained for these

independent variables is also less than the significant level of 0.05. Also the standard error values

are nearer to 0 which is good as we can see in summary of fit6 model

Though the value of adjusted R-squared value is less 0.05 the f-stat value is higher than rest of

the models that is 11.72.

furthermore when we observe the QQ plot for residuals they don't form any pattern

model fit6 is the best fit model for multiple regression analysis on our data, It can be

further proved by doing stepAIC on the fit1 model having all independent variables which

gives least AIC value for variables with fit6 model

If we predict the value of rating using energy= 86, dB = -3 and duration =120 then we get

predicted value of rating as 4

adjusted r square value can be improved by doing logistic analysis further