

Homework 3

R Markdown

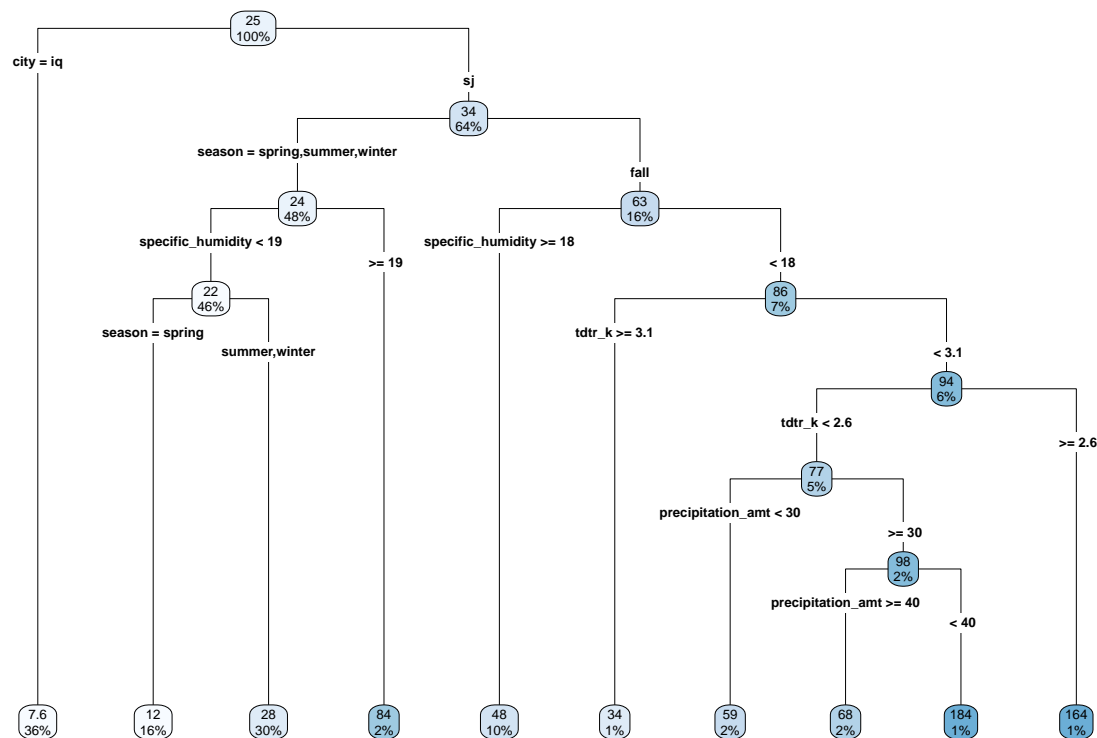
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The reason why is that different towns may have demographic differences. One town may have extremely poor residents, who are just more likely to turn to a life of crime, despite a higher presence of police. If we do not control for other variables, they could exert an effect on the dependent variable that is not correlated with our independent variable of interest. They used the instrumental variable of the event where there is heightened police presence without a crime rate-related trigger. They discovered that the “high alert” variable did exert a negative effect on crime occurrence. The control for “metro ridership” is to control for a proxy for population throughput in the area. We can reasonably assume that the more people in the area, the more likely a crime will be committed due to the increased presence of opportunity. The model estimated here places a differential on the effect of being on high alert on crime rate based on the geographic area in DC.

```
# CART
dengue_split = initial_split(dengue, prop = 0.8)
dengue_train = training(dengue_split)
dengue_test = testing(dengue_split)

dengue_tree = rpart(total_cases ~ city + season + specific_humidity + tdtr_k + precipitation_amt, data=
rpart.plot(dengue_tree, type=4)
```



```
print(dengue_tree)
```

```
## n= 1456
##
## node), split, n, deviance, yval
## * denotes terminal node
##
## 1) root 1456 2765389.000 24.675140
## 2) city=iq 520 60149.780 7.565385 *
## 3) city=sj 936 2468442.000 34.180560
## 6) season=spring,summer,winter 702 753319.000 24.420230
## 12) specific_humidity< 18.53929 676 440566.700 22.110950
## 24) season=spring 232 23842.000 11.500000 *
## 25) season=summer,winter 444 376954.300 27.655410 *
## 13) specific_humidity>=18.53929 26 215418.500 84.461540 *
## 7) season=fall 234 1447622.000 63.461540
## 14) specific_humidity>=17.65714 139 250747.400 48.107910 *
## 15) specific_humidity< 17.65714 95 1116164.000 85.926320
## 30) tdtr_k>=3.071429 13 5509.692 33.846150 *
## 31) tdtr_k< 3.071429 82 1069804.000 94.182930
## 62) tdtr_k< 2.592857 66 586233.100 77.272730
## 124) precipitation_amt< 29.84 35 176320.300 59.142860 *
## 125) precipitation_amt>=29.84 31 385419.900 97.741940
## 250) precipitation_amt>=39.64 23 137732.400 67.739130 *
## 251) precipitation_amt< 39.64 8 167460.000 184.000000 *
## 63) tdtr_k>=2.592857 16 386846.900 163.937500 *
```

```
summary(dengue_tree)
```

```
## Call:
## rpart(formula = total_cases ~ city + season + specific_humidity +
```

```

##      tdtr_k + precipitation_amt, data = dengue)
##      n= 1456
##
##      CP nsplit rel error      xerror      xstd
## 1 0.09118036      0 1.0000000 1.0014057 0.1625460
## 2 0.03519717      2 0.8176393 0.8268290 0.1337546
## 3 0.02918587      3 0.7824421 0.8544935 0.1331805
## 4 0.02487439      4 0.7532562 0.8428738 0.1290559
## 5 0.01893411      6 0.7035075 0.8528178 0.1322453
## 6 0.01438148      8 0.6656392 0.8496371 0.1306825
## 7 0.01000000      9 0.6512578 0.8386561 0.1292312
##
## Variable importance
##      tdtr_k      season      city specific_humidity
##      29      24      18      17
## precipitation_amt
##      12
##
## Node number 1: 1456 observations,      complexity param=0.09118036
##      mean=24.67514, MSE=1899.306
##      left son=2 (520 obs) right son=3 (936 obs)
##      Primary splits:
##      city      splits as LR,      improve=0.08562884, (0 missing)
##      tdtr_k      < 3.807143 to the right, improve=0.08456586, (10 missing)
##      season      splits as RLLL,      improve=0.07499072, (0 missing)
##      specific_humidity < 16.17429 to the left, improve=0.02231077, (10 missing)
##      precipitation_amt < 39.875 to the right, improve=0.00425222, (13 missing)
##      Surrogate splits:
##      tdtr_k      < 4.378571 to the right, agree=0.991, adj=0.975, (0 split)
##      precipitation_amt < 46.19 to the right, agree=0.700, adj=0.160, (0 split)
##      specific_humidity < 19.06143 to the right, agree=0.652, adj=0.027, (0 split)
##
## Node number 2: 520 observations
##      mean=7.565385, MSE=115.6726
##
## Node number 3: 936 observations,      complexity param=0.09118036
##      mean=34.18056, MSE=2637.225
##      left son=6 (702 obs) right son=7 (234 obs)
##      Primary splits:
##      season      splits as RLLL,      improve=0.108368500, (0 missing)
##      specific_humidity < 16.145 to the left, improve=0.046732810, (6 missing)
##      precipitation_amt < 16.01 to the left, improve=0.017904020, (9 missing)
##      tdtr_k      < 1.807143 to the right, improve=0.009452844, (6 missing)
##      Surrogate splits:
##      specific_humidity < 18.07214 to the left, agree=0.76, adj=0.038, (0 split)
##
## Node number 6: 702 observations,      complexity param=0.03519717
##      mean=24.42023, MSE=1073.104
##      left son=12 (676 obs) right son=13 (26 obs)
##      Primary splits:
##      specific_humidity < 18.53929 to the left, improve=0.12994630, (6 missing)
##      season      splits as -LRR,      improve=0.07598967, (0 missing)
##      tdtr_k      < 1.892857 to the right, improve=0.01550045, (6 missing)
##      precipitation_amt < 15.375 to the left, improve=0.00824373, (9 missing)

```

```

##
## Node number 7: 234 observations,      complexity param=0.02918587
##   mean=63.46154, MSE=6186.419
##   left son=14 (139 obs) right son=15 (95 obs)
##   Primary splits:
##     specific_humidity < 17.65714 to the right, improve=0.05575370, (0 missing)
##     tdtr_k             < 1.778571 to the right, improve=0.01985307, (0 missing)
##     precipitation_amt < 5.75      to the right, improve=0.01286678, (0 missing)
##   Surrogate splits:
##     precipitation_amt < 20.22     to the right, agree=0.667, adj=0.179, (0 split)
##     tdtr_k            < 1.664286 to the right, agree=0.607, adj=0.032, (0 split)
##
## Node number 12: 676 observations,      complexity param=0.01438148
##   mean=22.11095, MSE=651.7259
##   left son=24 (232 obs) right son=25 (444 obs)
##   Primary splits:
##     season              splits as  -LRR,          improve=0.090271020, (0 missing)
##     tdtr_k              < 2.535714 to the right, improve=0.025131630, (6 missing)
##     specific_humidity < 18.26571 to the left,  improve=0.015093980, (6 missing)
##     precipitation_amt < 45.29     to the left,  improve=0.007513767, (9 missing)
##   Surrogate splits:
##     tdtr_k < 2.992857 to the right, agree=0.683, adj=0.078, (0 split)
##
## Node number 13: 26 observations
##   mean=84.46154, MSE=8285.325
##
## Node number 14: 139 observations
##   mean=48.10791, MSE=1803.938
##
## Node number 15: 95 observations,      complexity param=0.02487439
##   mean=85.92632, MSE=11749.1
##   left son=30 (13 obs) right son=31 (82 obs)
##   Primary splits:
##     tdtr_k              < 3.071429 to the right, improve=0.03659903, (0 missing)
##     specific_humidity < 16.71071 to the left,  improve=0.03385087, (0 missing)
##     precipitation_amt < 16.59     to the left,  improve=0.01315308, (0 missing)
##   Surrogate splits:
##     specific_humidity < 15.43714 to the left,  agree=0.874, adj=0.077, (0 split)
##
## Node number 24: 232 observations
##   mean=11.5, MSE=102.7672
##
## Node number 25: 444 observations
##   mean=27.65541, MSE=848.9961
##
## Node number 30: 13 observations
##   mean=33.84615, MSE=423.8225
##
## Node number 31: 82 observations,      complexity param=0.02487439
##   mean=94.18293, MSE=13046.39
##   left son=62 (66 obs) right son=63 (16 obs)
##   Primary splits:
##     tdtr_k              < 2.592857 to the left, improve=0.09041301, (0 missing)
##     specific_humidity < 16.84643 to the left,  improve=0.03487083, (0 missing)

```

```

##      precipitation_amt < 0.315      to the left,  improve=0.01509650, (0 missing)
##
## Node number 62: 66 observations,      complexity param=0.01893411
##   mean=77.27273, MSE=8882.32
##   left son=124 (35 obs) right son=125 (31 obs)
##   Primary splits:
##     precipitation_amt < 29.84      to the left,  improve=0.04178009, (0 missing)
##     specific_humidity < 17.26071 to the left,  improve=0.04099777, (0 missing)
##     tdr_k              < 1.85      to the right, improve=0.03896871, (0 missing)
##   Surrogate splits:
##     tdr_k              < 2.007143 to the right, agree=0.697, adj=0.355, (0 split)
##     specific_humidity < 17.30143 to the left,  agree=0.667, adj=0.290, (0 split)
##
## Node number 63: 16 observations
##   mean=163.9375, MSE=24177.93
##
## Node number 124: 35 observations
##   mean=59.14286, MSE=5037.722
##
## Node number 125: 31 observations,      complexity param=0.01893411
##   mean=97.74194, MSE=12432.9
##   left son=250 (23 obs) right son=251 (8 obs)
##   Primary splits:
##     precipitation_amt < 39.64      to the right, improve=0.20815610, (0 missing)
##     specific_humidity < 16.91357 to the left,  improve=0.08434850, (0 missing)
##     tdr_k              < 1.864286 to the right, improve=0.02292126, (0 missing)
##   Surrogate splits:
##     specific_humidity < 16.40214 to the right, agree=0.774, adj=0.125, (0 split)
##
## Node number 250: 23 observations
##   mean=67.73913, MSE=5988.367
##
## Node number 251: 8 observations
##   mean=184, MSE=20932.5

```

```
predict (dengue_tree, newdata=dengue)
```

##	1	2	3	4	5	6	7
##	11.500000	11.500000	11.500000	11.500000	11.500000	27.655405	27.655405
##	8	9	10	11	12	13	14
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	15	16	17	18	19	20	21
##	27.655405	27.655405	27.655405	27.655405	33.846154	48.107914	48.107914
##	22	23	24	25	26	27	28
##	67.739130	48.107914	48.107914	48.107914	48.107914	48.107914	48.107914
##	29	30	31	32	33	34	35
##	59.142857	59.142857	184.000000	27.655405	27.655405	27.655405	27.655405
##	36	37	38	39	40	41	42
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	43	44	45	46	47	48	49
##	27.655405	27.655405	11.500000	11.500000	11.500000	11.500000	11.500000
##	50	51	52	53	54	55	56
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	57	58	59	60	61	62	63
##	11.500000	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405

##	64	65	66	67	68	69	70
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	71	72	73	74	75	76	77
##	48.107914	48.107914	48.107914	48.107914	48.107914	184.000000	59.142857
##	78	79	80	81	82	83	84
##	163.937500	163.937500	163.937500	67.739130	59.142857	184.000000	27.655405
##	85	86	87	88	89	90	91
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	92	93	94	95	96	97	98
##	27.655405	27.655405	27.655405	27.655405	27.655405	11.500000	11.500000
##	99	100	101	102	103	104	105
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	106	107	108	109	110	111	112
##	11.500000	11.500000	11.500000	11.500000	27.655405	27.655405	27.655405
##	113	114	115	116	117	118	119
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	120	121	122	123	124	125	126
##	27.655405	27.655405	27.655405	67.739130	59.142857	67.739130	48.107914
##	127	128	129	130	131	132	133
##	67.739130	48.107914	163.937500	59.142857	163.937500	48.107914	67.739130
##	134	135	136	137	138	139	140
##	67.739130	67.739130	27.655405	27.655405	27.655405	27.655405	27.655405
##	141	142	143	144	145	146	147
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	148	149	150	151	152	153	154
##	27.655405	27.655405	11.500000	11.500000	11.500000	11.500000	11.500000
##	155	156	157	158	159	160	161
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	162	163	164	165	166	167	168
##	11.500000	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	169	170	171	172	173	174	175
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	176	177	178	179	180	181	182
##	67.739130	48.107914	48.107914	48.107914	48.107914	48.107914	59.142857
##	183	184	185	186	187	188	189
##	67.739130	59.142857	59.142857	59.142857	67.739130	59.142857	27.655405
##	190	191	192	193	194	195	196
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	197	198	199	200	201	202	203
##	27.655405	27.655405	27.655405	27.655405	27.655405	11.500000	11.500000
##	204	205	206	207	208	209	210
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	211	212	213	214	215	216	217
##	11.500000	11.500000	11.500000	11.500000	27.655405	27.655405	27.655405
##	218	219	220	221	222	223	224
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	225	226	227	228	229	230	231
##	27.655405	27.655405	27.655405	163.937500	67.739130	184.000000	163.937500
##	232	233	234	235	236	237	238
##	163.937500	163.937500	163.937500	67.739130	59.142857	184.000000	184.000000
##	239	240	241	242	243	244	245
##	48.107914	59.142857	27.655405	27.655405	27.655405	27.655405	27.655405
##	246	247	248	249	250	251	252
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405

##	253	254	255	256	257	258	259
##	27.655405	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	260	261	262	263	264	265	266
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	267	268	269	270	271	272	273
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	274	275	276	277	278	279	280
##	84.461538	84.461538	27.655405	84.461538	27.655405	27.655405	48.107914
##	281	282	283	284	285	286	287
##	48.107914	48.107914	48.107914	48.107914	48.107914	67.739130	48.107914
##	288	289	290	291	292	293	294
##	48.107914	59.142857	59.142857	67.739130	59.142857	27.655405	27.655405
##	295	296	297	298	299	300	301
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	302	303	304	305	306	307	308
##	27.655405	27.655405	27.655405	11.500000	11.500000	11.500000	11.500000
##	309	310	311	312	313	314	315
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	316	317	318	319	320	321	322
##	11.500000	11.500000	27.655405	27.655405	27.655405	27.655405	27.655405
##	323	324	325	326	327	328	329
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	330	331	332	333	334	335	336
##	27.655405	48.107914	48.107914	48.107914	48.107914	59.142857	48.107914
##	337	338	339	340	341	342	343
##	59.142857	48.107914	59.142857	48.107914	184.000000	67.739130	67.739130
##	344	345	346	347	348	349	350
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	351	352	353	354	355	356	357
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	11.500000
##	358	359	360	361	362	363	364
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	365	366	367	368	369	370	371
##	11.500000	11.500000	11.500000	11.500000	11.500000	27.655405	27.655405
##	372	373	374	375	376	377	378
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	379	380	381	382	383	384	385
##	27.655405	27.655405	27.655405	27.655405	48.107914	48.107914	48.107914
##	386	387	388	389	390	391	392
##	48.107914	48.107914	48.107914	48.107914	48.107914	163.937500	59.142857
##	393	394	395	396	397	398	399
##	59.142857	48.107914	48.107914	27.655405	27.655405	27.655405	27.655405
##	400	401	402	403	404	405	406
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	407	408	409	410	411	412	413
##	27.655405	27.655405	11.500000	11.500000	11.500000	11.500000	11.500000
##	414	415	416	417	418	419	420
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	84.461538
##	421	422	423	424	425	426	427
##	11.500000	27.655405	84.461538	27.655405	27.655405	27.655405	27.655405
##	428	429	430	431	432	433	434
##	27.655405	27.655405	84.461538	84.461538	84.461538	84.461538	84.461538
##	435	436	437	438	439	440	441
##	48.107914	48.107914	48.107914	48.107914	48.107914	48.107914	48.107914

##	442	443	444	445	446	447	448
##	48.107914	48.107914	48.107914	48.107914	67.739130	59.142857	27.655405
##	449	450	451	452	453	454	455
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	456	457	458	459	460	461	462
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	11.500000
##	463	464	465	466	467	468	469
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	470	471	472	473	474	475	476
##	11.500000	11.500000	11.500000	11.500000	11.500000	27.655405	27.655405
##	477	478	479	480	481	482	483
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	84.461538
##	484	485	486	487	488	489	490
##	27.655405	27.655405	27.655405	27.655405	48.107914	48.107914	48.107914
##	491	492	493	494	495	496	497
##	48.107914	48.107914	67.739130	48.107914	48.107914	48.107914	59.142857
##	498	499	500	501	502	503	504
##	48.107914	48.107914	33.846154	27.655405	27.655405	27.655405	27.655405
##	505	506	507	508	509	510	511
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	512	513	514	515	516	517	518
##	27.655405	27.655405	11.500000	11.500000	11.500000	11.500000	11.500000
##	519	520	521	522	523	524	525
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	526	527	528	529	530	531	532
##	11.500000	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	533	534	535	536	537	538	539
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	540	541	542	543	544	545	546
##	48.107914	48.107914	48.107914	48.107914	48.107914	33.846154	48.107914
##	547	548	549	550	551	552	553
##	33.846154	48.107914	33.846154	33.846154	163.937500	59.142857	27.655405
##	554	555	556	557	558	559	560
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	561	562	563	564	565	566	567
##	27.655405	27.655405	27.655405	27.655405	11.500000	11.500000	11.500000
##	568	569	570	571	572	573	574
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	575	576	577	578	579	580	581
##	11.500000	11.500000	11.500000	27.655405	27.655405	27.655405	27.655405
##	582	583	584	585	586	587	588
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	84.461538
##	589	590	591	592	593	594	595
##	84.461538	27.655405	48.107914	48.107914	48.107914	48.107914	48.107914
##	596	597	598	599	600	601	602
##	48.107914	48.107914	48.107914	48.107914	33.846154	67.739130	33.846154
##	603	604	605	606	607	608	609
##	33.846154	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	610	611	612	613	614	615	616
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	617	618	619	620	621	622	623
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	624	625	626	627	628	629	630
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	27.655405

##	631	632	633	634	635	636	637
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	638	639	640	641	642	643	644
##	27.655405	27.655405	27.655405	27.655405	27.655405	48.107914	48.107914
##	645	646	647	648	649	650	651
##	48.107914	48.107914	48.107914	48.107914	48.107914	48.107914	48.107914
##	652	653	654	655	656	657	658
##	33.846154	163.937500	163.937500	59.142857	27.655405	27.655405	27.655405
##	659	660	661	662	663	664	665
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	666	667	668	669	670	671	672
##	27.655405	27.655405	27.655405	11.500000	11.500000	11.500000	11.500000
##	673	674	675	676	677	678	679
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	680	681	682	683	684	685	686
##	11.500000	11.500000	27.655405	27.655405	27.655405	27.655405	27.655405
##	687	688	689	690	691	692	693
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	694	695	696	697	698	699	700
##	84.461538	48.107914	48.107914	48.107914	48.107914	48.107914	48.107914
##	701	702	703	704	705	706	707
##	48.107914	48.107914	48.107914	33.846154	48.107914	48.107914	59.142857
##	708	709	710	711	712	713	714
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	715	716	717	718	719	720	721
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	11.500000
##	722	723	724	725	726	727	728
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##	729	730	731	732	733	734	735
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##	736	737	738	739	740	741	742
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	743	744	745	746	747	748	749
##	27.655405	27.655405	84.461538	27.655405	48.107914	48.107914	48.107914
##	750	751	752	753	754	755	756
##	48.107914	48.107914	48.107914	48.107914	48.107914	48.107914	67.739130
##	757	758	759	760	761	762	763
##	59.142857	67.739130	59.142857	27.655405	27.655405	27.655405	27.655405
##	764	765	766	767	768	769	770
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	771	772	773	774	775	776	777
##	27.655405	27.655405	27.655405	11.500000	11.500000	11.500000	11.500000
##	778	779	780	781	782	783	784
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	785	786	787	788	789	790	791
##	11.500000	84.461538	27.655405	27.655405	84.461538	84.461538	84.461538
##	792	793	794	795	796	797	798
##	27.655405	27.655405	84.461538	84.461538	84.461538	84.461538	27.655405
##	799	800	801	802	803	804	805
##	84.461538	48.107914	48.107914	48.107914	48.107914	48.107914	48.107914
##	806	807	808	809	810	811	812
##	48.107914	48.107914	48.107914	48.107914	184.000000	59.142857	59.142857
##	813	814	815	816	817	818	819
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405

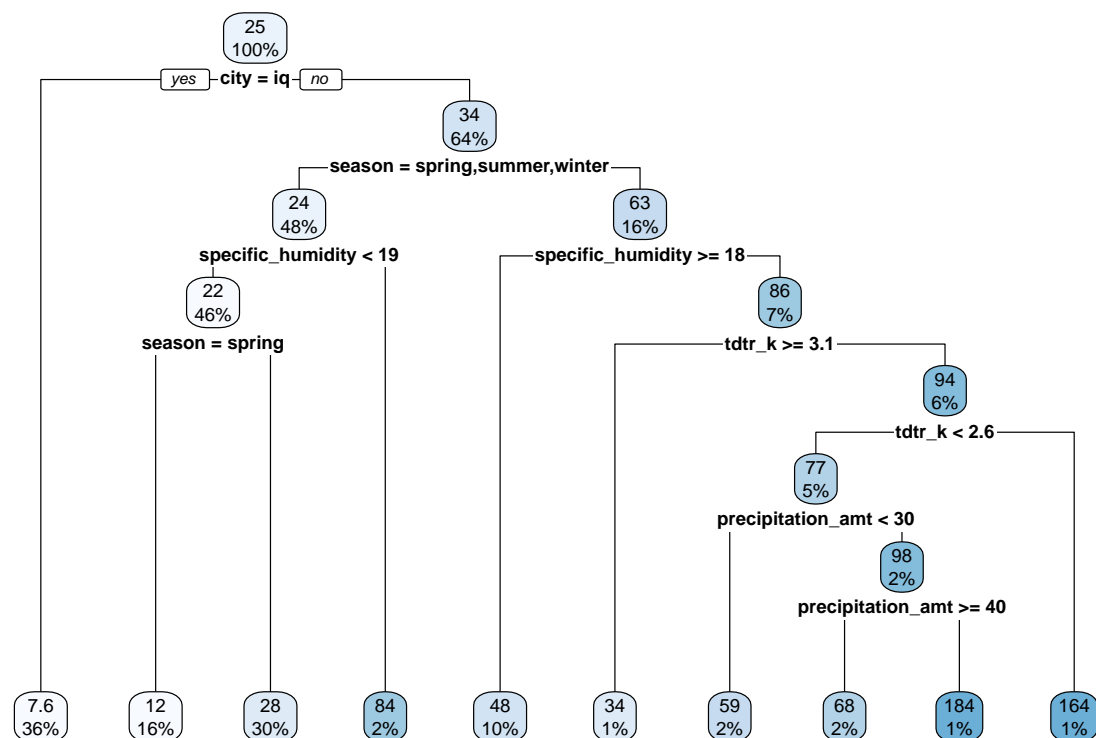
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##	827	828	829	830	831	832	833
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	834	835	836	837	838	839	840
##	11.500000	11.500000	11.500000	11.500000	11.500000	27.655405	27.655405
##	841	842	843	844	845	846	847
##	27.655405	84.461538	27.655405	27.655405	27.655405	27.655405	27.655405
##	848	849	850	851	852	853	854
##	27.655405	27.655405	27.655405	84.461538	48.107914	48.107914	48.107914
##	855	856	857	858	859	860	861
##	48.107914	48.107914	48.107914	48.107914	48.107914	48.107914	59.142857
##	862	863	864	865	866	867	868
##	59.142857	33.846154	163.937500	27.655405	27.655405	27.655405	27.655405
##	869	870	871	872	873	874	875
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##	876	877	878	879	880	881	882
##	27.655405	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	883	884	885	886	887	888	889
##	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000	11.500000
##	890	891	892	893	894	895	896
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405
##	897	898	899	900	901	902	903
##	27.655405	27.655405	27.655405	27.655405	27.655405	27.655405	48.107914
##	904	905	906	907	908	909	910
##	48.107914	48.107914	48.107914	48.107914	48.107914	48.107914	59.142857
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##	163.937500	59.142857	33.846154	67.739130	59.142857	27.655405	27.655405
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##	960	961	962	963	964	965	966
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##	988	989	990	991	992	993	994
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##	995	996	997	998	999	1000	1001
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##	1002	1003	1004	1005	1006	1007	1008
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##	1009	1010	1011	1012	1013	1014	1015
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##	1016	1017	1018	1019	1020	1021	1022
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##	1030	1031	1032	1033	1034	1035	1036
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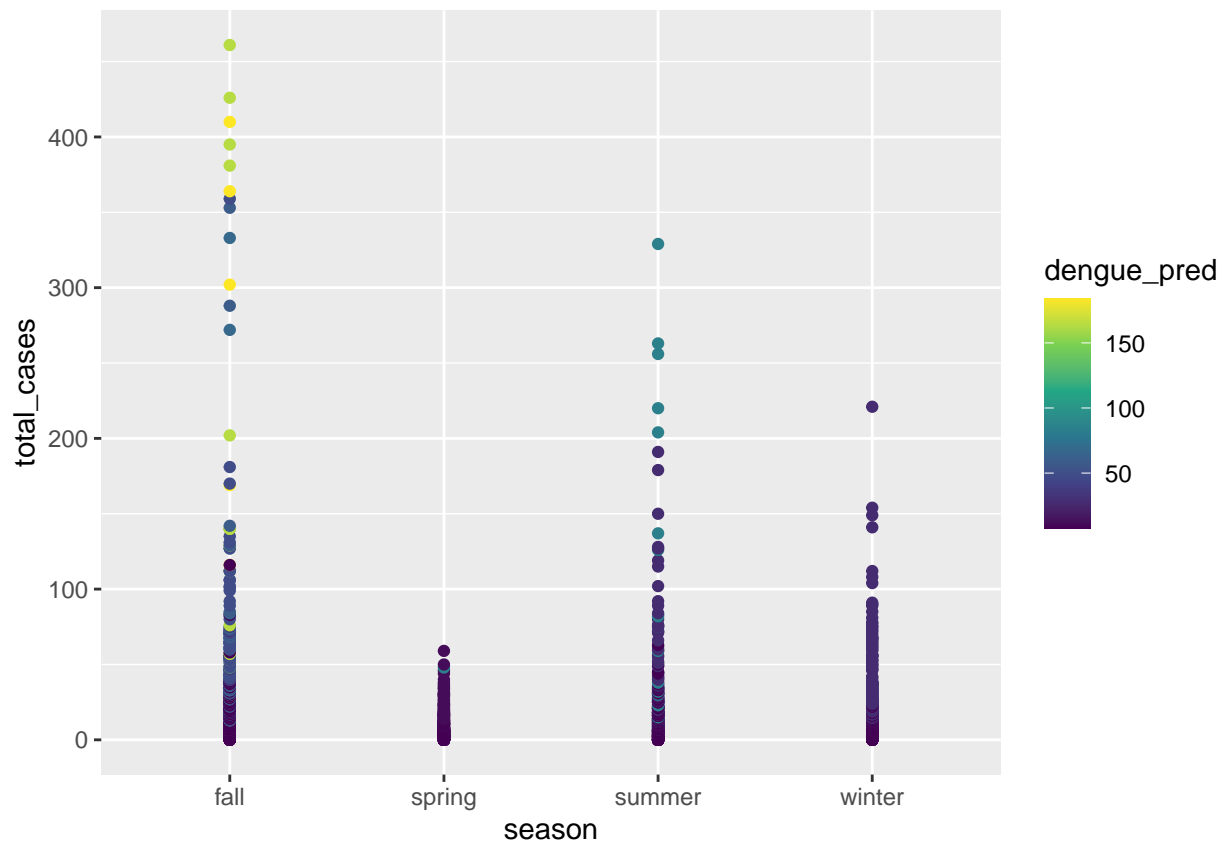
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##      1401      1402      1403      1404      1405      1406      1407
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##      1408      1409      1410      1411      1412      1413      1414
##  7.565385  7.565385  7.565385  7.565385  7.565385  7.565385  7.565385
##      1415      1416      1417      1418      1419      1420      1421
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##      1422      1423      1424      1425      1426      1427      1428
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##      1429      1430      1431      1432      1433      1434      1435
##  7.565385  7.565385  7.565385  7.565385  7.565385  7.565385  7.565385
##      1436      1437      1438      1439      1440      1441      1442
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##      1443      1444      1445      1446      1447      1448      1449
##  7.565385  7.565385  7.565385  7.565385  7.565385  7.565385  7.565385
##      1450      1451      1452      1453      1454      1455      1456
##  7.565385  7.565385  7.565385  7.565385  7.565385  7.565385  7.565385
```

```
load.tree = rpart(total_cases ~ city + season + specific_humidity + tdr_k + precipitation_amt, data =
rpart.plot(load.tree))
```

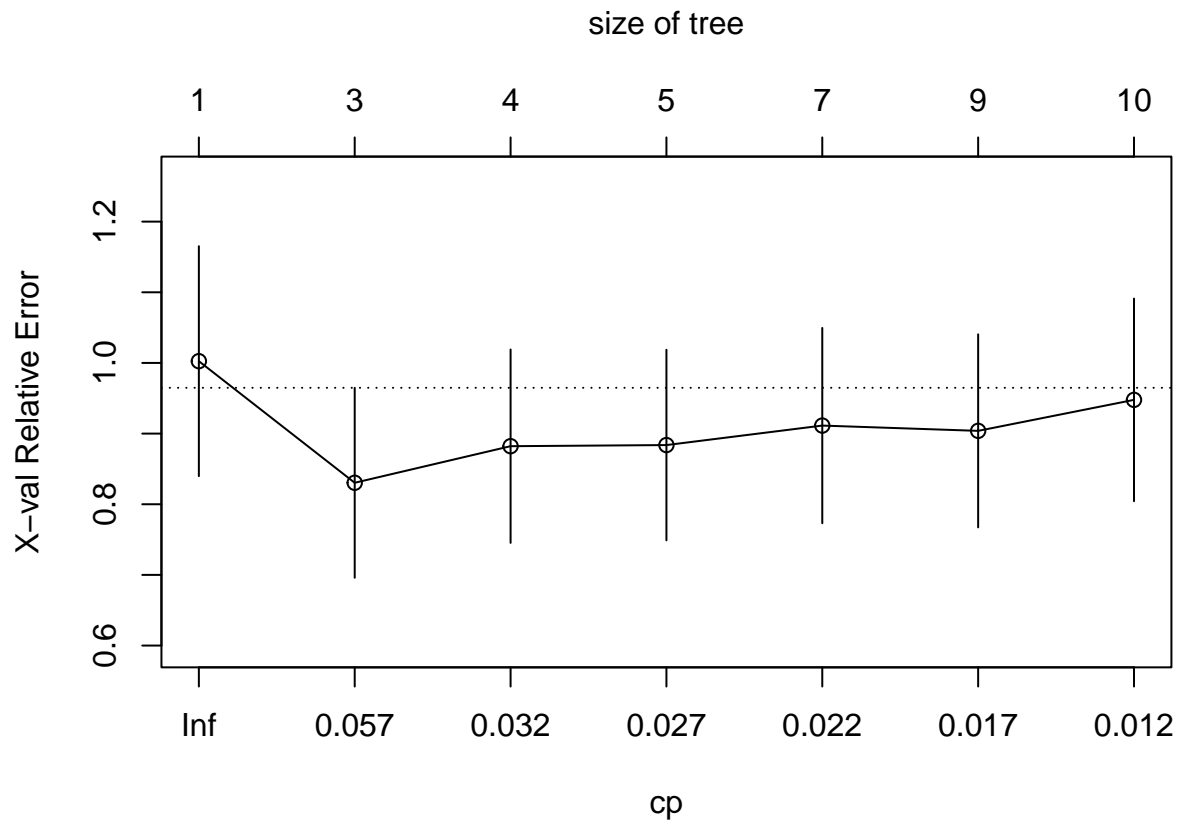


```
reg_dengue = reg_dengue %>%
  mutate (dengue_pred = predict(load.tree)) %>%
  arrange(total_cases)

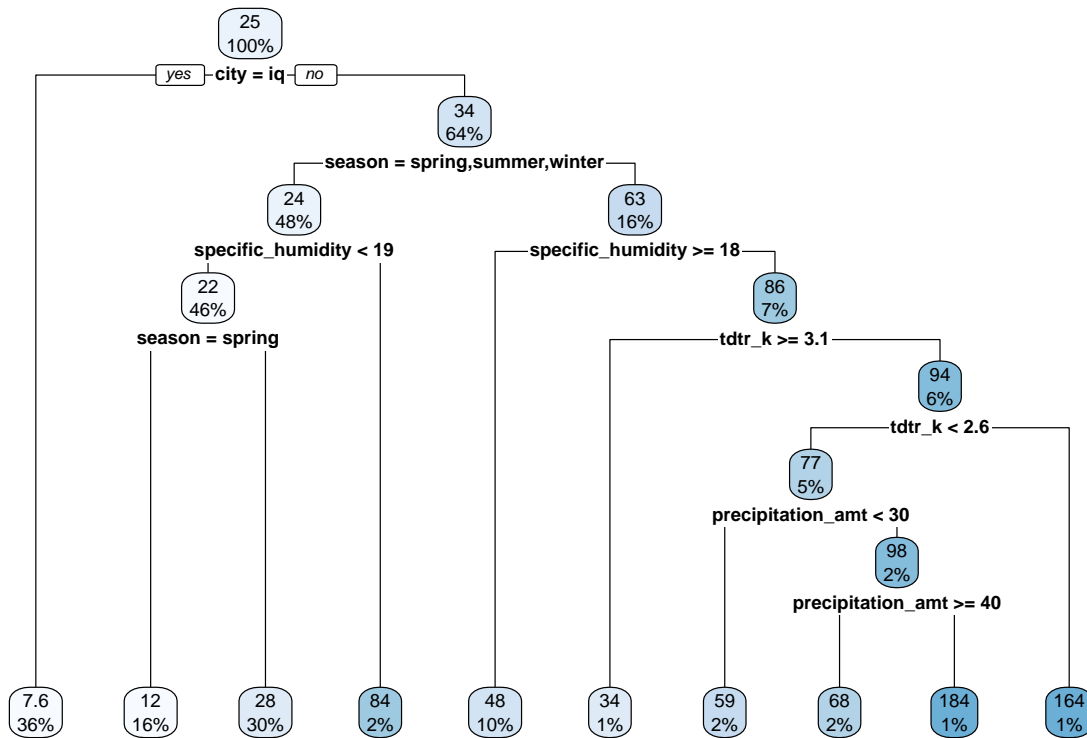
ggplot(reg_dengue) +
  geom_point(aes(x=season, y=total_cases, color=dengue_pred)) +
  scale_color_continuous(type = "viridis")
```



```
plotcp(load.tree)
```



```
rpart.plot(load.tree)
```



```
printcp(load.tree)
```

```
##
## Regression tree:
## rpart(formula = total_cases ~ city + season + specific_humidity +
##       tdtr_k + precipitation_amt, data = reg_dengue)
##
## Variables actually used in tree construction:
## [1] city          precipitation_amt season          specific_humidity
## [5] tdtr_k
##
## Root node error: 2765389/1456 = 1899.3
##
## n= 1456
##
##      CP nsplit rel error  xerror   xstd
## 1 0.091180      0  1.00000 1.00251 0.16275
## 2 0.035197      2  0.81764 0.83037 0.13445
## 3 0.029186      3  0.78244 0.88218 0.13689
## 4 0.024874      4  0.75326 0.88376 0.13485
## 5 0.018934      6  0.70351 0.91134 0.13824
## 6 0.014381      8  0.66564 0.90388 0.13662
## 7 0.010000      9  0.65126 0.94772 0.14340
```

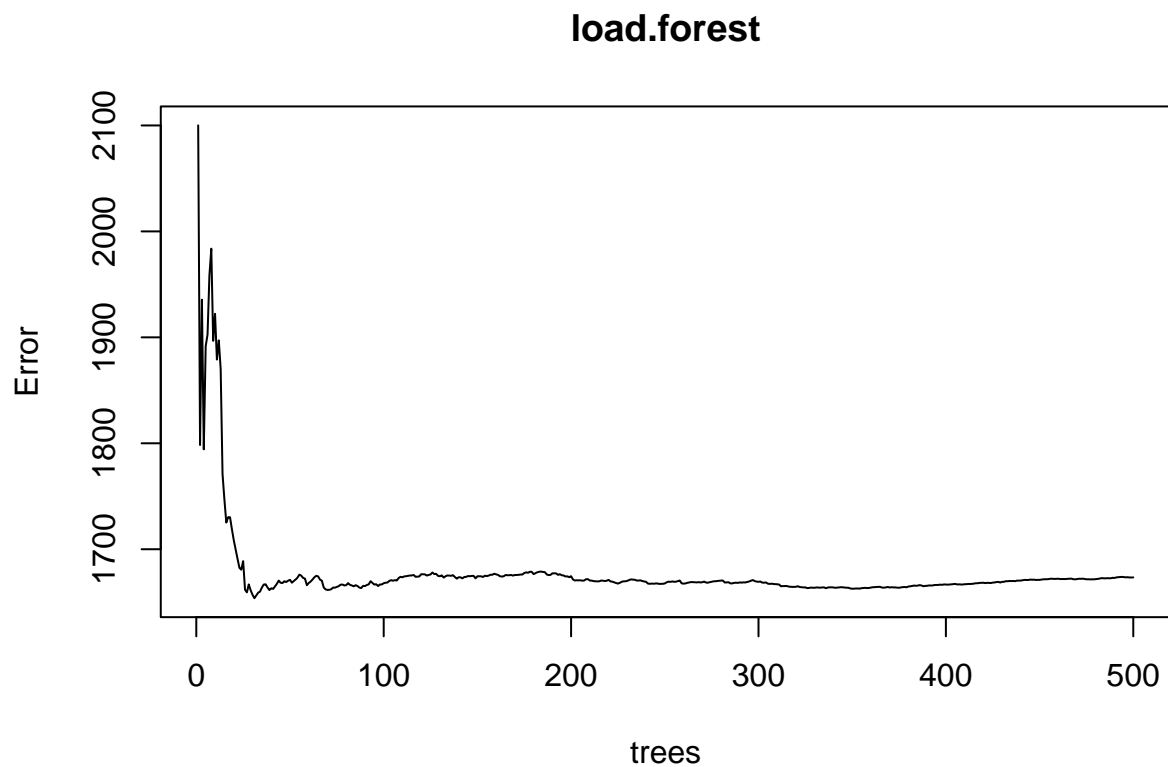
#Random Forrest

```
load_tree2= read.csv("https://raw.githubusercontent.com/jgscott/EC0395M/master/data/dengue.csv")
head(load_tree2)
```

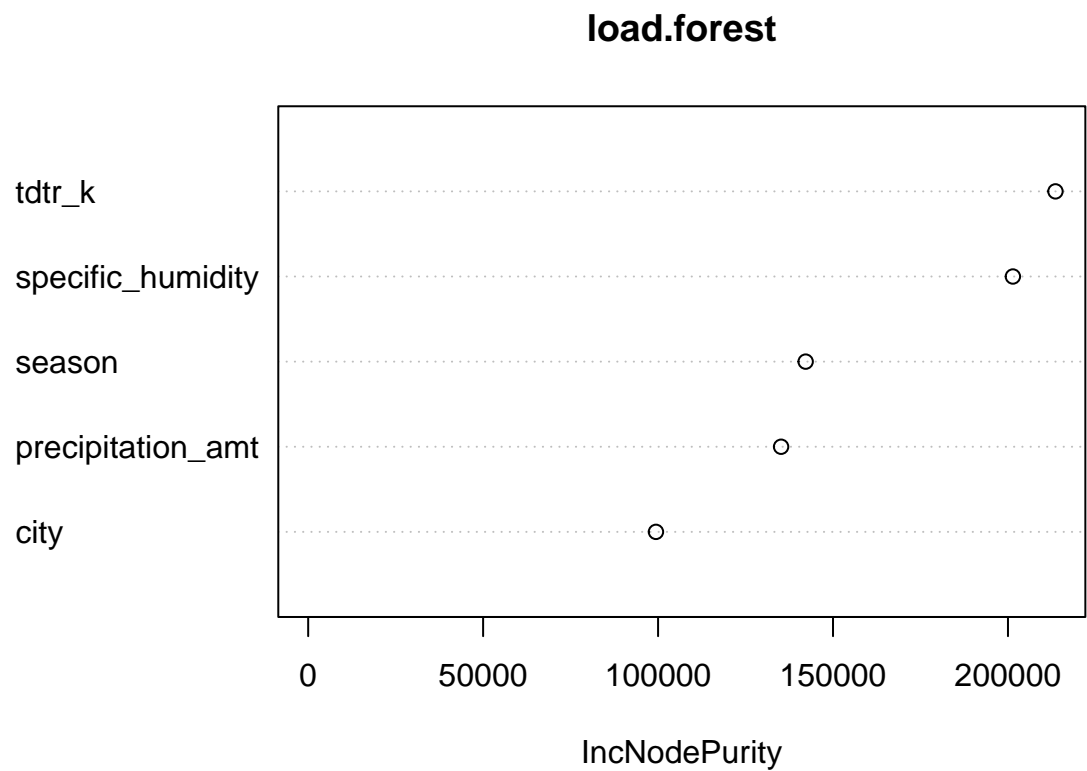
```
##   city season total_cases   ndvi_ne   ndvi_nw   ndvi_se   ndvi_sw
## 1   sj spring           4 0.1226000 0.1037250 0.1984833 0.1776167
## 2   sj spring           5 0.1699000 0.1421750 0.1623571 0.1554857
## 3   sj spring           4 0.0322500 0.1729667 0.1572000 0.1708429
## 4   sj spring           3 0.1286333 0.2450667 0.2275571 0.2358857
## 5   sj spring           6 0.1962000 0.2622000 0.2512000 0.2473400
## 6   sj summer           2      NA 0.1748500 0.2543143 0.1817429
##   precipitation_amt air_temp_k avg_temp_k dew_point_temp_k max_air_temp_k
## 1                12.42   297.5729   297.7429           292.4143         299.8
## 2                22.82   298.2114   298.4429           293.9514         300.9
## 3                34.54   298.7814   298.8786           295.4343         300.5
## 4                15.36   298.9871   299.2286           295.3100         301.4
## 5                 7.52   299.5186   299.6643           295.8214         301.9
## 6                 9.58   299.6300   299.7643           295.8514         302.4
##   min_air_temp_k precip_amt_kg_per_m2 relative_humidity_percent
## 1             295.9                32.00             73.36571
## 2             296.4                17.94             77.36857
## 3             297.3                26.10             82.05286
## 4             297.0                13.90             80.33714
## 5             297.5                12.20             80.46000
## 6             298.1                26.49             79.89143
##   specific_humidity   tdtr_k
## 1          14.01286 2.628571
## 2          15.37286 2.371429
## 3          16.84857 2.300000
## 4          16.67286 2.428571
## 5          17.21000 3.014286
## 6          17.21286 2.100000
```

```
load_split = initial_split(load_tree2, prop=0.8)
load_train = training(load_split)
load_test  = testing(load_split)
```

```
load.forest = randomForest(total_cases ~ city + season + specific_humidity + tdtr_k + precipitation_amt
plot(load.forest)
```

```
vi = varImpPlot(load.forest, type=2)
```



```
par(mfrow=c(3,3))
nms = names(load_train)[1:9]
for(i in 1:9) plot(load.forest,i=nms[i])
```

```

## Warning in plot.window(...): "i" is not a graphical parameter
## Warning in plot.xy(xy, type, ...): "i" is not a graphical parameter
## Warning in axis(side = side, at = at, labels = labels, ...): "i" is not a
## graphical parameter

## Warning in axis(side = side, at = at, labels = labels, ...): "i" is not a
## graphical parameter
## Warning in box(...): "i" is not a graphical parameter
## Warning in title(...): "i" is not a graphical parameter
## Warning in plot.window(...): "i" is not a graphical parameter
## Warning in plot.xy(xy, type, ...): "i" is not a graphical parameter
## Warning in axis(side = side, at = at, labels = labels, ...): "i" is not a
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## graphical parameter

## Warning in axis(side = side, at = at, labels = labels, ...): "i" is not a
## graphical parameter

```

```

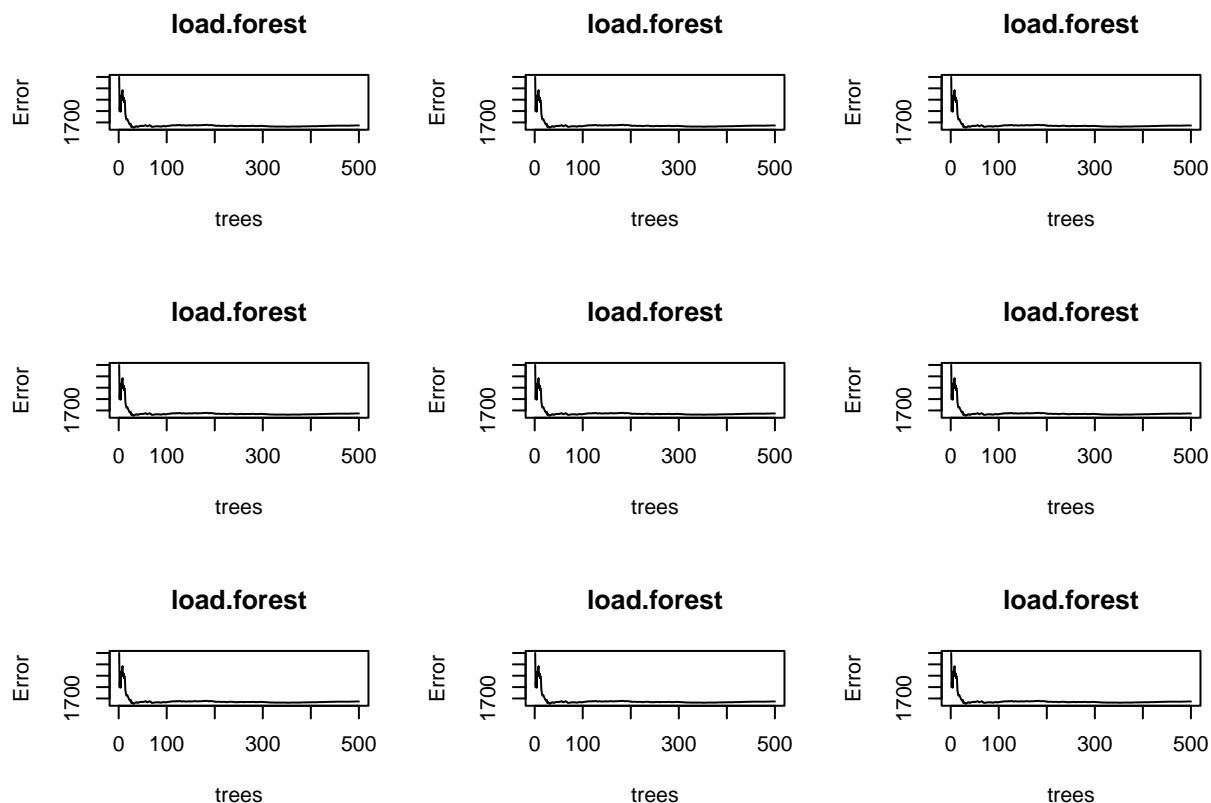
## Warning in box(...): "i" is not a graphical parameter
## Warning in title(...): "i" is not a graphical parameter
## Warning in plot.window(...): "i" is not a graphical parameter
## Warning in plot.xy(xy, type, ...): "i" is not a graphical parameter
## Warning in axis(side = side, at = at, labels = labels, ...): "i" is not a
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## Warning in axis(side = side, at = at, labels = labels, ...): "i" is not a
## graphical parameter

## Warning in axis(side = side, at = at, labels = labels, ...): "i" is not a
## graphical parameter
## Warning in box(...): "i" is not a graphical parameter
## Warning in title(...): "i" is not a graphical parameter

```



```
boost1 = gbm(total_cases ~ factor(city) + factor(season) + specific_humidity + tdtr_k + precipitation_at_location,
              interaction.depth=2, n.trees=300, shrinkage=.05)
```

```
## Distribution not specified, assuming gaussian ...
```

```
gbm.perf(boost1)
```

```
## OOB generally underestimates the optimal number of iterations although predictive performance is reasonably good
```

```
## [1] 31
```

```
## attr(,"smoother")
```

```
## Call:
```

```
## loess(formula = object$oobag.improve ~ x, enp.target = min(max(4,
##   length(x)/10), 50))
```

```
##
```

```
## Number of Observations: 300
```

```
## Equivalent Number of Parameters: 24.11
```

```
## Residual Standard Error: 1.54
```

```
modelr::rmse(load.forest, load_test)
```

```
## [1] 32.77927
```

```
modelr::rmse(load.tree, load_test)
```

```
## [1] 33.46296
```

```
modelr::rmse(boost1,load_test)
```

```
## Using 300 trees...
```

```
## [1] 34.30779
```

```
gbm.perf(boost1)
```

```
## OOB generally underestimates the optimal number of iterations although predictive performance is rea
```

```
## [1] 31
```

```
## attr("smoother")
```

```
## Call:
```

```
## loess(formula = object$oobag.improve ~ x, enp.target = min(max(4,  
##     length(x)/10), 50))
```

```
##
```

```
## Number of Observations: 300
```

```
## Equivalent Number of Parameters: 24.11
```

```
## Residual Standard Error: 1.54
```

```
# The Cart Model has the lowest RMSE, we will construct PDP based on it
```

```
load.tree %>%
```

```
  partial(pred.var = "specific_humidity" ) %>% autoplot
```

```
## Warning: Use of `object[[1L]]` is discouraged. Use `.data[[1L]]` instead.
```

```
## Warning: Use of `object[["yhat"]]` is discouraged. Use `.data[["yhat"]]`  
## instead.
```

```
load.tree %>%
```

```
  partial(pred.var = "precipitation_amt" ) %>% autoplot
```

```
## Warning: Use of `object[[1L]]` is discouraged. Use `.data[[1L]]` instead.
```

```
## Warning: Use of `object[["yhat"]]` is discouraged. Use `.data[["yhat"]]`  
## instead.
```

```
load.tree %>%
```

```
  partial(pred.var = "tdtr_k" ) %>% autoplot
```

```
## Warning: Use of `object[[1L]]` is discouraged. Use `.data[[1L]]` instead.
```

```
## Warning: Use of `object[["yhat"]]` is discouraged. Use `.data[["yhat"]]`  
## instead.
```

```
#Green Building
```

```
green = green %>%
```

```
  mutate(green_cert = ifelse( LEED == 1 | Energystar == 1,  
                              yes=1, no=0))%>%
```

```
  mutate(revenue = Rent* leasing_rate/100)%>%
```

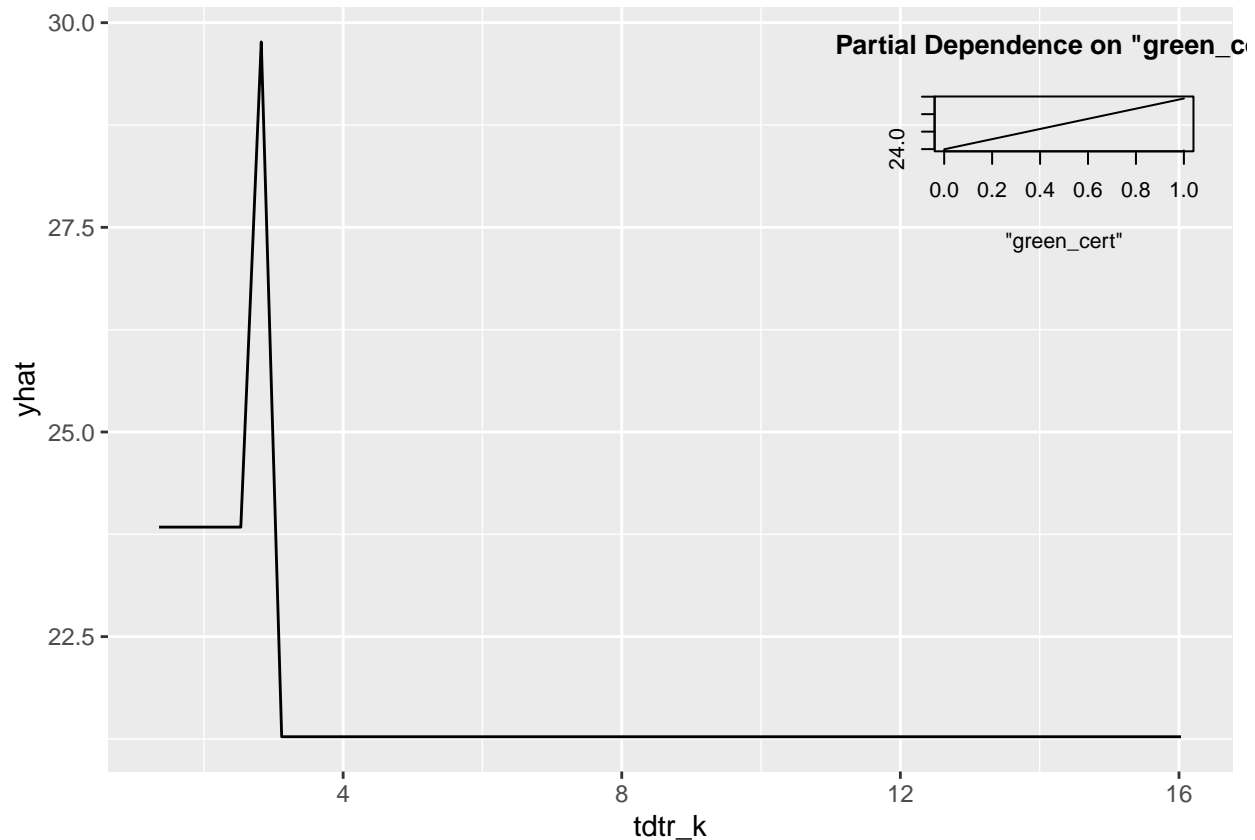
```
  mutate(lrevenue = log(revenue))
```

```
forest_green = randomForest(revenue ~ class_a + stories + size + cluster + age + renovated + green_cert
```

```
pd_green = partialPlot(forest_green,green, "green_cert")
```

```
print(pd_green)
```

```
## $x
## [1] 0 1
##
## $y
## [1] 23.99526 25.44774
```



Report for Green Building.

By using a random forest model, I fitted a predicative model for the revenue per square foot per calendar that took in to considerations the age, stories, size and location of the building, as well as the build quality. By building the partial dependence plot, I was able to isolated the effect of having a green certificated (LEED or Energy Star), to about \$1.3 dollars increase in revenue.

You can also embed plots, for example:

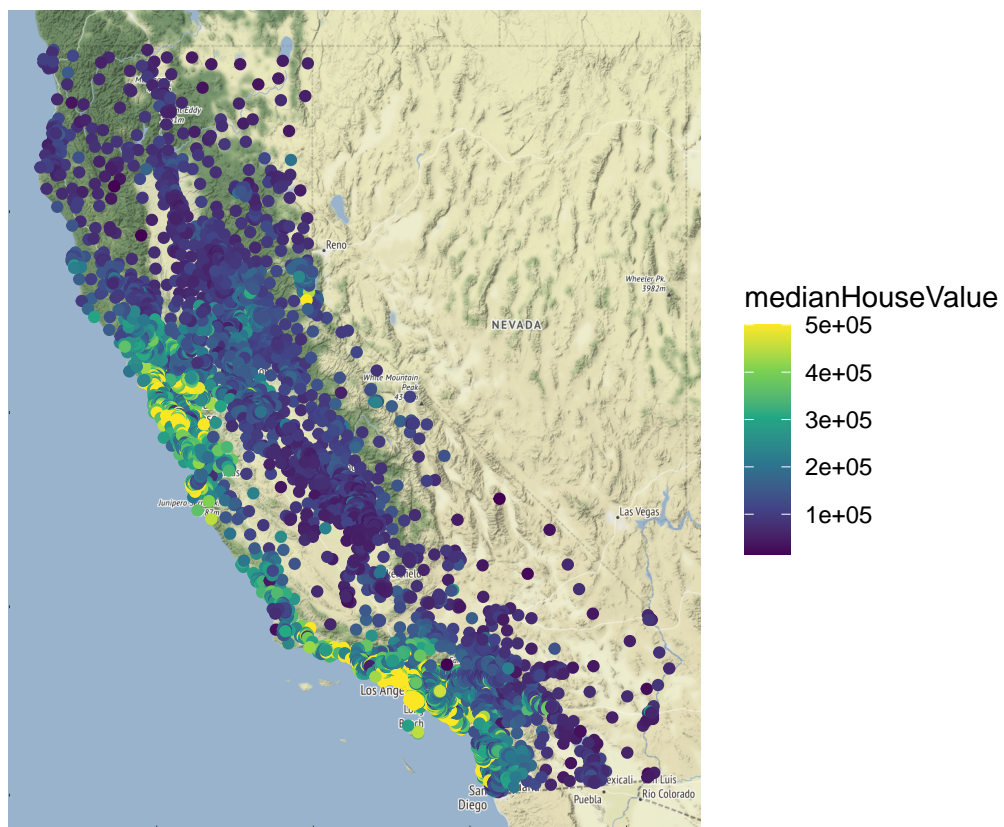
```
## # A tibble: 6 x 9
##   longitude latitude housingMedianAge totalRooms totalBedrooms population
##   <dbl>     <dbl>         <dbl>     <dbl>         <dbl>         <dbl>
## 1   -122.     37.9           41        880           129           322
## 2   -122.     37.9           21       7099          1106          2401
## 3   -122.     37.8           52       1467           190           496
## 4   -122.     37.8           52       1274           235           558
## 5   -122.     37.8           52       1627           280           565
## 6   -122.     37.8           52        919           213           413
## # ... with 3 more variables: households <dbl>, medianIncome <dbl>,
## #   medianHouseValue <dbl>

## Using zoom = 7...

## Source : http://tile.stamen.com/terrain/7/19/47.png
```

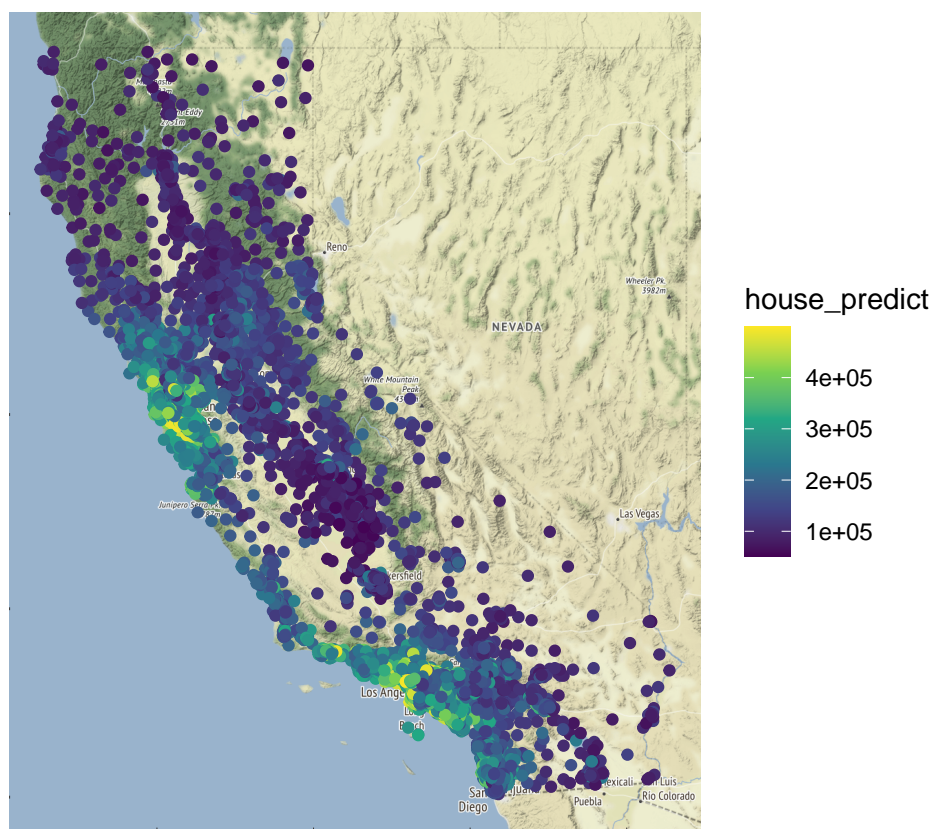
Source : <http://tile.stamen.com/terrain/7/20/47.png>
Source : <http://tile.stamen.com/terrain/7/21/47.png>
Source : <http://tile.stamen.com/terrain/7/22/47.png>
Source : <http://tile.stamen.com/terrain/7/23/47.png>
Source : <http://tile.stamen.com/terrain/7/19/48.png>
Source : <http://tile.stamen.com/terrain/7/20/48.png>
Source : <http://tile.stamen.com/terrain/7/21/48.png>
Source : <http://tile.stamen.com/terrain/7/22/48.png>
Source : <http://tile.stamen.com/terrain/7/23/48.png>
Source : <http://tile.stamen.com/terrain/7/19/49.png>
Source : <http://tile.stamen.com/terrain/7/20/49.png>
Source : <http://tile.stamen.com/terrain/7/21/49.png>
Source : <http://tile.stamen.com/terrain/7/22/49.png>
Source : <http://tile.stamen.com/terrain/7/23/49.png>
Source : <http://tile.stamen.com/terrain/7/19/50.png>
Source : <http://tile.stamen.com/terrain/7/20/50.png>
Source : <http://tile.stamen.com/terrain/7/21/50.png>
Source : <http://tile.stamen.com/terrain/7/22/50.png>
Source : <http://tile.stamen.com/terrain/7/23/50.png>
Source : <http://tile.stamen.com/terrain/7/19/51.png>
Source : <http://tile.stamen.com/terrain/7/20/51.png>
Source : <http://tile.stamen.com/terrain/7/21/51.png>
Source : <http://tile.stamen.com/terrain/7/22/51.png>
Source : <http://tile.stamen.com/terrain/7/23/51.png>

Original Data



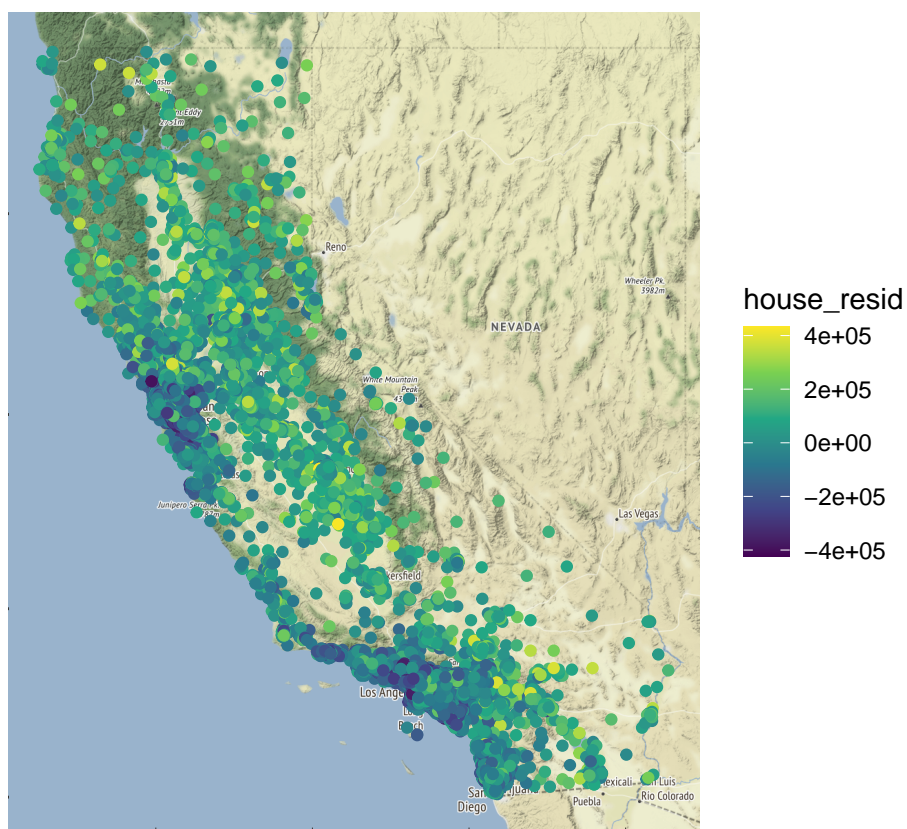
Using zoom = 7...

Model Predication



Using zoom = 7...

Residuals



```
## [1] 51405.59
```

Note that the `echo = FALSE` parameter was added to the code chunk to prevent printing of the R code that generated the plot.