

DPA Project

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```
library(readxl)

WaterQualityData=data.frame()

for (x in 2017:2021) {
  path="C:/Users/SRINU/Desktop/DPA Project/WQ_"
  file=as.character(x)
  filetype=".xlsx"
  finalPath=paste(path,file,filetype,sep="")
  DataByYear=read_excel(finalPath)
  DataByYear=DataByYear[-1:-2,]
  DataByYear=data.frame(DataByYear,x)
  WaterQualityData=rbind(WaterQualityData,DataByYear)
}

colnames(WaterQualityData) =c('StateCode','StationName','State','TemperatureMin','TemperatureMax','DissolvedSolids')

library(readxl)
library(dplyr)

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
## 
##     filter, lag

## The following objects are masked from 'package:base':
## 
##     intersect, setdiff, setequal, union

library(zoo)

##
## Attaching package: 'zoo'

## The following objects are masked from 'package:base':
## 
##     as.Date, as.Date.numeric
```

```

AirQualityDataFrame=data.frame()

for (x in 2017:2019) {
  path="C:/Users/SRINU/Desktop/DPA Project/AQ_"
  file=as.character(x)
  filetype=".xlsx"
  # Specify the Excel file path
  file_path = paste(path,file,filetype,sep="")

  # Read all sheet names from the Excel file
  sheet_names = excel_sheets(file_path)

  # Create an empty list to store data frames
  all_sheets = list()

  # Loop through all sheet names
  for (sheet_name in sheet_names) {
    # Read the sheet into a data frame
    sheet_df = read_excel(file_path, sheet = sheet_name, range = NULL)

    # Check if the sheet is not blank (i.e., has rows and columns)
    if (!is.null(dim(sheet_df))) {
      measure=sheet_df[1,]

      sheet_df$measure=paste(na.omit(unlist(measure)), collapse = ", ")
      sheet_df = sheet_df[-c(1, 2), ]
      sheet_df = na.locf(sheet_df, na.rm = FALSE)
      sheet_df$year=as.character(x)

      #colnames(sheet_df)=c('State','City','Location','MonitoringDays','Min','Max','Average','measure','year')
      # Add the data frame to the list of all sheets
      all_sheets[[sheet_name]] = sheet_df
    }
  }

  # Combine all non-blank sheets into a single data frame
  merged_df = bind_rows(all_sheets)
}

AirQualityDataFrame=rbind(AirQualityDataFrame,merged_df)

}

colnames(AirQualityDataFrame)=c('State','City','Location','MonitoringDays','Min','Max','Average','measure','year')

library(dplyr)

AirQualityDataFrame = AirQualityDataFrame %>%
  mutate(measure = case_when(
    measure == "PM_{2.5} Concentration in pg/m^{3}" ~ "PM2.5",
    measure == "PM_{10} Concentration in pg/m^{3}" ~ "PM10",
    measure == "NO_{2} Concentration in pg/m^{3}" ~ "NO2",

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measure == "SO_{2} Concentration in µg/m^{3}" ~ "SO2",
measure == "State, SO_{2} Concentration in µg/m^{3}" ~ "SO2",
measure == "PM_{10} concentration µg/m^{3}" ~ "PM10",
measure == "NO_{2} concentration in µg/m^{3}" ~ "NO2",
measure == "PM_{2.5} concentration µg/m^{3}" ~ "PM2.5"
))

library(dplyr)

# Convert Average column to numeric data type
AirQualityDataFrame$Average = as.integer(AirQualityDataFrame$Average)

# Aggregate data by year, state, city, and measure
agg_AirQualityDataFrame = AirQualityDataFrame %>%
  group_by(year, State, City, measure) %>%
  summarize(Mean = mean(Average), .groups = "drop")

library(tidyr)

# Pivot the Type column into new columns
agg_AirQualityDataFrame = pivot_wider(agg_AirQualityDataFrame, names_from = measure, values_from = Mean)

agg_AirQualityDataFrames=agg_AirQualityDataFrame[agg_AirQualityDataFrame$year==2017,]

library(dplyr)

# Group the data by year and calculate the mean of the PM10 column for each year
means_by_year = agg_AirQualityDataFrames %>%
  group_by(year) %>%
  summarize(mean_PM10 = mean(PM10, na.rm = TRUE))

# Join the means back to the original data frame by year
agg_AirQualityDataFrame_filled = left_join(agg_AirQualityDataFrames, means_by_year, by = "year")

# Fill missing PM10s in the PM10 column with the mean PM10 by year
agg_AirQualityDataFrame_filled$PM10[is.na(agg_AirQualityDataFrame_filled$PM10)] = agg_AirQualityDataFrame_filled$mean_PM10

means_by_year = agg_AirQualityDataFrame_filled %>%
  group_by(year) %>%
  summarize(mean_NO2 = mean(NO2, na.rm = TRUE))

# Join the means back to the original data frame by year
agg_AirQualityDataFrame_filled = left_join(agg_AirQualityDataFrame_filled, means_by_year, by = "year")

# Fill missing NO2s in the NO2 column with the mean NO2 by year
agg_AirQualityDataFrame_filled$NO2[is.na(agg_AirQualityDataFrame_filled$NO2)] = agg_AirQualityDataFrame_filled$mean_NO2

means_by_year = agg_AirQualityDataFrame_filled %>%

```

```

group_by(year) %>%
  summarize(mean_SO2 = mean(SO2, na.rm = TRUE))

# Join the means back to the original data frame by year
agg_AirQualityDataFrame_filled = left_join(agg_AirQualityDataFrame_filled, means_by_year, by = "year")

# Fill missing SO2s in the SO2 column with the mean SO2 by year
agg_AirQualityDataFrame_filled$SO2[is.na(agg_AirQualityDataFrame_filled$SO2)] = agg_AirQualityDataFrame_filled$mean_SO2

means_by_year = agg_AirQualityDataFrame_filled %>%
  group_by(year) %>%
  summarize(mean_PM2.5 = mean(PM2.5, na.rm = TRUE))

# Join the means back to the original data frame by year
agg_AirQualityDataFrame_filled = left_join(agg_AirQualityDataFrame_filled, means_by_year, by = "year")

# Fill missing PM2.5s in the PM2.5 column with the mean PM2.5 by year
agg_AirQualityDataFrame_filled$PM2.5[is.na(agg_AirQualityDataFrame_filled$PM2.5)] = agg_AirQualityDataFrame_filled$mean_PM2.5

AirQualityDF=agg_AirQualityDataFrame_filled[,c(-8,-9,-10,-11)]

```

```

library(readxl)

AQ_2021 = read_excel("C:/Users/SRINU/Desktop/DPA Project/AQ_2021.xlsx")

AQ_2021=AQ_2021[-1:-2,]

AQ_2021$year='2021'

AQ_2021 = AQ_2021[, c(7,1,2,3,4,5,6)]

AQ_2021[AQ_2021 == "-"] = NA

colnames(AQ_2021)=c('year','State','City','SO2','PM2.5','NO2','PM10')

AQ_2021$SO2      =as.numeric(AQ_2021$SO2)
AQ_2021$PM2.5    =as.numeric(AQ_2021$PM2.5)
AQ_2021$NO2      =as.numeric(AQ_2021$PM2.5)
AQ_2021$PM10     =as.numeric(AQ_2021$PM10)

AQ_2021$SO2[is.na(AQ_2021$SO2)] = mean(AQ_2021$SO2, na.rm = TRUE)
AQ_2021$PM10[is.na(AQ_2021$PM10)] = mean(AQ_2021$PM10, na.rm = TRUE)
AQ_2021$PM2.5[is.na(AQ_2021$PM2.5)] = mean(AQ_2021$PM2.5, na.rm = TRUE)
AQ_2021$NO2[is.na(AQ_2021$NO2)] = mean(AQ_2021$NO2, na.rm = TRUE)

AQ_2021 = AQ_2021[, c("year","State","City","NO2","PM10","SO2","PM2.5")]

AirQualityDataset = rbind(AirQualityDF, AQ_2021)

```

```
library(stringr)

View(WaterQualityData)

WaterQualityData$StationName = ifelse(grepl("AT", WaterQualityData$StationName), sub(".*AT\\s+(\\w+).*", 
WaterQualityData$StationName = ifelse(grepl(" ", WaterQualityData$StationName), sub("^.+\\s*(\\S+)$", " 

names(WaterQualityData) [2] = "City"

WaterQualityData

##           StateCode
## 1            1448
## 2            2352
## 3            1393
## 4            2459
## 5            2460
## 6            2461
## 7            2462
## 8            2463
## 9            2464
## 10           2465
## 11           2466
## 12           1399
## 13           1400
## 14           1475
## 15           1476
## 16           1543
## 17           1544
## 18           1545
## 19           1546
## 20           1547
## 21           1548
## 22           2270
## 23           2271
## 24           2272
## 25           2273
## 26           2274
## 27           2275
## 28           2276
## 29           3181
## 30           3182
## 31           3183
## 32           3184
## 33           3190
## 34           3191
## 35           1148
## 36           1149
## 37           1150
```

## 38	1246
## 39	1434
## 40	1435
## 41	1436
## 42	1438
## 43	1860
## 44	1861
## 45	1862
## 46	1865
## 47	1980
## 48	2072
## 49	2082
## 50	3204
## 51	3205
## 52	1025
## 53	1026
## 54	1884
## 55	1885
## 56	1887
## 57	1871
## 58	2624
## 59	2625
## 60	3862
## 61	3863
## 62	3864
## 63	1444
## 64	1892
## 65	1894
## 66	2780
## 67	3562
## 68	3563
## 69	17
## 70	18
## 71	20
## 72	21
## 73	42
## 74	43
## 75	1154
## 76	1155
## 77	1156
## 78	1338
## 79	1339
## 80	1340
## 81	1341
## 82	1342
## 83	1384
## 84	1442
## 85	1443
## 86	1563
## 87	1564
## 88	1565
## 89	1566
## 90	1567
## 91	1568

## 92	1569
## 93	1570
## 94	1571
## 95	1572
## 96	1573
## 97	2284
## 98	2285
## 99	2286
## 100	2287
## 101	2288
## 102	2289
## 103	2290
## 104	2291
## 105	2292
## 106	2293
## 107	2294
## 108	2295
## 109	2296
## 110	2297
## 111	2298
## 112	2299
## 113	2300
## 114	2301
## 115	2302
## 116	2303
## 117	2304
## 118	2305
## 119	2306
## 120	2307
## 121	2319
## 122	2326
## 123	2331
## 124	2332
## 125	2333
## 126	2334
## 127	2335
## 128	2336
## 129	2337
## 130	2338
## 131	3458
## 132	3459
## 133	3460
## 134	3464
## 135	3465
## 136	3466
## 137	3468
## 138	3471
## 139	3473
## 140	1092
## 141	1093
## 142	1094
## 143	1151
## 144	1152
## 145	1461

## 146	1462
## 147	2162
## 148	2164
## 149	2168
## 150	2198
## 151	2199
## 152	2651
## 153	2653
## 154	2654
## 155	2670
## 156	2671
## 157	2672
## 158	2676
## 159	2685
## 160	2686
## 161	2687
## 162	2688
## 163	2689
## 164	2696
## 165	2701
## 166	2702
## 167	2703
## 168	2704
## 169	2706
## 170	2707
## 171	2708
## 172	2709
## 173	2712
## 174	2713
## 175	2714
## 176	1424
## 177	1457
## 178	1458
## 179	1624
## 180	1625
## 181	1626
## 182	1627
## 183	1628
## 184	1925
## 185	1926
## 186	1927
## 187	2852
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## 189	2856
## 190	2857
## 191	2858
## 192	2859
## 193	2860
## 194	2861
## 195	2862
## 196	2863
## 197	2864
## 198	2867
## 199	2868

## 200	2870
## 201	2872
## 202	2873
## 203	2874
## 204	2876
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## 206	2878
## 207	2879
## 208	2880
## 209	1427
## 210	1428
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## 212	1632
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## 230	3366
## 231	3375
## 232	3376
## 233	3379
## 234	3383
## 235	3384
## 236	2442
## 237	2443
## 238	2444
## 239	2445
## 240	1023
## 241	1024
## 242	1295
## 243	1473
## 244	1698
## 245	1699
## 246	1700
## 247	1701
## 248	1702
## 249	1703
## 250	1704
## 251	1705
## 252	2914
## 253	2947

## 254	2948
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## 256	1159
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## 264	3026
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## 267	3029
## 268	1403
## 269	1404
## 270	1726
## 271	3397
## 272	3398
## 273	3399
## 274	3400
## 275	3401
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## 278	3404
## 279	3405
## 280	3406
## 281	3407
## 282	3408
## 283	3409
## 284	3410
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## 289	3415
## 290	3416
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## 293	3422
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## 505	1023
## 506	1024
## 507	1295
## 508	1473
## 509	1698
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## 514	1703
## 515	1704
## 516	1705
## 517	2914
## 518	1717
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## 521	2947
## 522	2948
## 523	2949

## 524	4171
## 525	1159
## 526	1160
## 527	1161
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## 538	1403
## 539	1404
## 540	1726
## 541	3397
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## 543	3399
## 544	3400
## 545	3401
## 546	3402
## 547	3403
## 548	3404
## 549	3405
## 550	3406
## 551	3407
## 552	3408
## 553	3409
## 554	3410
## 555	3411
## 556	3412
## 557	3413
## 558	3414
## 559	3415
## 560	3416
## 561	3420
## 562	3421
## 563	3422
## 564	4358
## 565	1448
## 566	4346
## 567	4347
## 568	4348
## 569	4351
## 570	4390
## 571	4357
## 572	4379
## 573	2352
## 574	4369
## 575	3760
## 576	3765
## 577	3785

## 578	3789
## 579	3820
## 580	3807
## 581	3821
## 582	3777
## 583	3763
## 584	3787
## 585	3788
## 586	3768
## 587	3791
## 588	3812
## 589	3799
## 590	3796
## 591	3810
## 592	3803
## 593	3783
## 594	3800
## 595	3782
## 596	3767
## 597	3808
## 598	3811
## 599	3798
## 600	3780
## 601	3814
## 602	3819
## 603	3774
## 604	3775
## 605	3784
## 606	3773
## 607	4405
## 608	4399
## 609	4400
## 610	4401
## 611	30017
## 612	30018
## 613	30020
## 614	30019
## 615	30008
## 616	1393
## 617	2459
## 618	2460
## 619	2461
## 620	2462
## 621	2463
## 622	2464
## 623	2465
## 624	2466
## 625	1548
## 626	2276
## 627	4410
## 628	2275
## 629	1543
## 630	1546
## 631	2270

## 632	2272
## 633	1545
## 634	1400
## 635	1476
## 636	10012
## 637	2274
## 638	2271
## 639	2273
## 640	3183
## 641	3184
## 642	4409
## 643	3190
## 644	3191
## 645	1547
## 646	1544
## 647	1399
## 648	1475
## 649	3181
## 650	3182
## 651	1148
## 652	1434
## 653	1980
## 654	1860
## 655	1436
## 656	1150
## 657	1246
## 658	1865
## 659	4420
## 660	1862
## 661	1149
## 662	1438
## 663	1435
## 664	1861
## 665	4423
## 666	2082
## 667	3204
## 668	3205
## 669	1025
## 670	1026
## 671	1884
## 672	1885
## 673	1887
## 674	4464
## 675	4465
## 676	4036
## 677	4023
## 678	4024
## 679	1871
## 680	2624
## 681	2625
## 682	3862
## 683	3863
## 684	3864
## 685	4025

## 686	4026
## 687	4038
## 688	4076
## 689	4079
## 690	4081
## 691	4742
## 692	4753
## 693	4754
## 694	4738
## 695	4739
## 696	4745
## 697	4747
## 698	4756
## 699	4499
## 700	4500
## 701	1444
## 702	2780
## 703	4497
## 704	1894
## 705	1892
## 706	3562
## 707	3563
## 708	1342
## 709	1443
## 710	3465
## 711	3466
## 712	2297
## 713	2319
## 714	2286
## 715	2331
## 716	2332
## 717	1154
## 718	20
## 719	21
## 720	3464
## 721	1572
## 722	1573
## 723	2295
## 724	1564
## 725	2287
## 726	1566
## 727	2293
## 728	2337
## 729	2338
## 730	42
## 731	2294
## 732	3460
## 733	3458
## 734	1571
## 735	1155
## 736	3471
## 737	2289
## 738	2301
## 739	2291

## 740	2326
## 741	1569
## 742	2298
## 743	1567
## 744	1568
## 745	2285
## 746	1340
## 747	1384
## 748	3459
## 749	2307
## 750	1339
## 751	2304
## 752	43
## 753	1570
## 754	2302
## 755	1563
## 756	2284
## 757	2288
## 758	1341
## 759	1565
## 760	10016
## 761	10017
## 762	1156
## 763	17
## 764	18
## 765	1338
## 766	2333
## 767	2334
## 768	2335
## 769	2336
## 770	3468
## 771	2300
## 772	2303
## 773	2290
## 774	2299
## 775	2305
## 776	2296
## 777	2292
## 778	2306
## 779	3473
## 780	1442
## 781	4620
## 782	4619
## 783	4611
## 784	4613
## 785	4624
## 786	4606
## 787	4621
## 788	4612
## 789	4609
## 790	2651
## 791	1461
## 792	2653
## 793	2654

## 794	1092
## 795	2670
## 796	1152
## 797	2198
## 798	2671
## 799	2672
## 800	2168
## 801	2676
## 802	1151
## 803	1462
## 804	2685
## 805	2686
## 806	2687
## 807	2688
## 808	2689
## 809	2696
## 810	2199
## 811	2701
## 812	2702
## 813	2703
## 814	2704
## 815	2706
## 816	2707
## 817	2708
## 818	2709
## 819	1093
## 820	1094
## 821	2162
## 822	2712
## 823	2164
## 824	2713
## 825	2714
## 826	1424
## 827	1457
## 828	1627
## 829	1628
## 830	2861
## 831	2862
## 832	2864
## 833	2874
## 834	1458
## 835	1624
## 836	2863
## 837	2873
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## 839	2877
## 840	1926
## 841	2878
## 842	2867
## 843	2852
## 844	2857
## 845	2879
## 846	1625
## 847	1626

## 848	2858
## 849	2859
## 850	2860
## 851	2880
## 852	1927
## 853	2856
## 854	2870
## 855	2872
## 856	2855
## 857	2876
## 858	3383
## 859	3384
## 860	1632
## 861	4112
## 862	1428
## 863	3375
## 864	3376
## 865	1631
## 866	3379
## 867	1633
## 868	4111
## 869	4326
## 870	3353
## 871	3354
## 872	3355
## 873	3356
## 874	3350
## 875	3351
## 876	3352
## 877	3359
## 878	3360
## 879	3361
## 880	3362
## 881	3357
## 882	3358
## 883	3363
## 884	3364
## 885	3365
## 886	1427
## 887	3366
## 888	2444
## 889	2442
## 890	2443
## 891	2445
## 892	1023
## 893	1024
## 894	1295
## 895	1473
## 896	1698
## 897	1699
## 898	1700
## 899	1701
## 900	1702
## 901	1703

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## 902      1704
## 903      1705
## 904      4159
## 905      4160
## 906      4162
## 907      2948
## 908      2947
## 909      2949
## 910      1450
## 911      1159
## 912      1160
## 913      1161
## 914      1162
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## 916      1329
## 917      1330
## 918      3025
## 919      3026
## 920      3027
## 921      3028
## 922      3029
## 923      10062
## 924      3405
## 925      3422
## 926      3414
## 927      3415
## 928      3416
## 929      3410
## 930      3411
## 931      1403
## 932      1404
## 933      3406
## 934      3407
## 935      3408
## 936      3409
## 937      1726
## 938      3397
## 939      3401
## 940      3402
## 941      3403
## 942      3421
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## 944      3399
## 945      3400
## 946      3404
## 947      3420
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## 949      3413
## 950  STN\r\nCode
## 951      1448
## 952      2352
## 953      4346
## 954      4347
## 955      4348
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## 956	4351
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## 958	4390
## 959	4922
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## 961	3763
## 962	3765
## 963	3767
## 964	3768
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## 967	3775
## 968	3777
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## 971	3783
## 972	3784
## 973	3786
## 974	3788
## 975	3791
## 976	3796
## 977	3799
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## 979	3807
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## 981	3811
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## 983	4400
## 984	4401
## 985	4405
## 986	1393
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## 991	2463
## 992	2464
## 993	2465
## 994	2466
## 995	30040
## 996	1399
## 997	1400
## 998	1475
## 999	1476
## 1000	1543
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## 1002	1545
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## 1004	1547
## 1005	1548
## 1006	2270
## 1007	2271
## 1008	2272
## 1009	2273

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## 1010      2274
## 1011      2275
## 1012      2276
## 1013      3181
## 1014      3182
## 1015      3183
## 1016      3184
## 1017      3190
## 1018      3191
## 1019      4409
## 1020      4410
## 1021     10012
## 1022      1148
## 1023      1149
## 1024      1150
## 1025      1246
## 1026      1434
## 1027      1435
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## 1029      1438
## 1030      1860
## 1031      1861
## 1032      1862
## 1033      1865
## 1034      1980
## 1035      2072
## 1036      2082
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## 1042      1884
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## 1046     30014
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## 1051      2624
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## 1053      3862
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## 1061      4036
## 1062      4037
## 1063      4038
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## 1067      4469
## 1068      4470
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## 1070      4479
## 1071      30008
## 1072      30013
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## 1075      4739
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## 1077      4745
## 1078      4747
## 1079      4753
## 1080      4754
## 1081      1444
## 1082      1892
## 1083      1894
## 1084      2780
## 1085      3562
## 1086      3563
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## 1091      4091
## 1092      4497
## 1093      4499
## 1094      4500
## 1095      17
## 1096      18
## 1097      20
## 1098      21
## 1099      42
## 1100      43
## 1101      1154
## 1102      1155
## 1103      1156
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## 1106      1340
## 1107      1341
## 1108      1342
## 1109      1384
## 1110      1442
## 1111      1443
## 1112      1563
## 1113      1564
## 1114      1565
## 1115      1566
## 1116      1567
## 1117      Dsf;s;dl
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## 1118      1569
## 1119      1570
## 1120      1571
## 1121      1572
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## 1148      2326
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## 1163      3468
## 1164      3471
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## 1166      10016
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## 1168      4606
## 1169      4609
## 1170      4611
## 1171      4612
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## 1180      1093
## 1181      1094
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## 1188      2168
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## 1200      2687
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## 1269	3376
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## 1271	3383
## 1272	3384
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## 1274	4112
## 1275	4326
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## 1278	2444
## 1279	2445

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## 1280      1023
## 1281      1024
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## 1288      1702
## 1289      1703
## 1290      1704
## 1291      1705
## 1292      4159
## 1293      4160
## 1294      4162
## 1295      30015
## 1296      30016
## 1297      30019
## 1298      2947
## 1299      2948
## 1300      2949
## 1301      4790
## 1302      1159
## 1303      1160
## 1304      1161
## 1305      1162
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## 1315      10062
## 1316      1403
## 1317      1404
## 1318      1726
## 1319      3397
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## 1321      3399
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## 1323      3401
## 1324      3402
## 1325      3403
## 1326      3404
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## 1330      3408
## 1331      3409
## 1332      3410
## 1333      3411
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## 1334      3412
## 1335      3413
## 1336      3414
## 1337      3415
## 1338      3416
## 1339      3420
## 1340      3421
## 1341      3422
## 1342 STN\r\nCode
## 1343      1448
## 1344      4351
## 1345      4346
## 1346      4347
## 1347      4348
## 1348      2352
## 1349      1393
## 1350      2461
## 1351      2460
## 1352      2463
## 1353      2465
## 1354      2462
## 1355      2466
## 1356      2459
## 1357      30040
## 1358      2464
## 1359      2276
## 1360      3189
## 1361      2275
## 1362      1543
## 1363      2270
## 1364      1546
## 1365      2272
## 1366      1545
## 1367      3185
## 1368      3186
## 1369      1400
## 1370      10012
## 1371      1476
## 1372      3187
## 1373      2274
## 1374      3183
## 1375      3184
## 1376      2273
## 1377      2271
## 1378      1547
## 1379      3188
## 1380      1544
## 1381      3181
## 1382      1399
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## 1384      1475
## 1385      2081
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## 1388      1980
## 1389      1860
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## 1398      1435
## 1399      1438
## 1400      1861
## 1401      2082
## 1402      30018
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## 1406      30017
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## 1409      1026
## 1410      30007
## 1411      30008
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## 1442      2338
## 1443        42
## 1444      2294
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## 1451      2291
## 1452      2326
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## 1461      2307
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## 1464        43
## 1465      2302
## 1466      1570
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## 1468      2284
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## 1472     10017
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## 1474     10016
## 1475        18
## 1476      2335
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## 1491      2306
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## 1496	1461
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## 1520	2162
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## 1522	1093
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## 1585      1427
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## 1592      3709
## 1593      2050
## 1594      4118
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## 1596      3717
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## 1600      2424
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## 1604      1642
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## 1617      2442
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## 1619      4160
## 1620      1024
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## 1658	4640	
## 1659	3397	
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## 1661	3401	
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## 1		RIVER SAL PAZORKHONI,\r\nCUNCOLIM(NEAR CULVER
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## 24		RIVER MAPUSA ON
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## 32		
## 33		RIVER SINQ
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## 35		
## 36		
## 37		
## 38		
## 39		AMLAKHADI AFTER
## 40		
## 41		BHADAR D/S JETPUR VILL. AFTER C
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BH
TRIVENI SANGAM, NR. SOMNATH

GHAGGAR BEFORE OTTU WEIR (BEF)

MARKANDA I

RAMPUR JATTAN MOGINAND NALA\r\nBEFORE CONFLUENCE TO RIVER\r\nMARKANDA
MA

KUMARADHARA - U/S OF UPPINAGADY\r\nTOWN BEFORE C

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KYRHUKHLA NEA

GAUTAMI-GOI
GAUTAI

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HAORA RIVER, D/S OF CHA

316 RIVER SAL PAZORKHONI
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319 RIVER MAPUSA ON\r\n## 320
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326 RIVER SINQUERIA
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333 AMLAKHADI AFTER C
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335 BHADAR D/S JETPUR V
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342 RIVER
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352 GHAGGAR BEFORE OTTU
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355 THUMBE WATER SUPPLIES
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PATALGA

KYRHUKHLA NEAR SUTNG

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MUNIGUDA (D/S C)

GAUTAMI-GO
GAUTAM

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564 RIVER GOWTHAMI (GODAVARI)\r\nGOVALANKA, INJARAM (V), NEAR G
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566 RIVER NAGAVALI NEAR NH- 16 BRIDGE,\r\nUPSTREAM (U/S) OF SRIKAKULAM\r\ntown,
567 RIVER NAGAVALI NEAR WATER PUMP\r\nHOUSE (KILLIPALEM) , DOWN STREAM\r\n(D/S) OF SRIKAKULAM TOWN
568 RIVER VAMSADHARA, IMMEDIATE\r\nBORDER BETWEEN ANDHRA PRADESH &\r\nORISSA STATES, NEAR LALITHAMB
569 RIVER NAGAVALI IMMEDIATE BORDER\r\nBETWEEN ANDHRA PRADESH & ORISSA\r\nSTATES, NEAR SIT
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574 COLLING WATER BLOWDOWNS FROM\r\nANDHRA SUGARS LTD, TAN
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KUJI RIVER (NEAR NEW BOGAGIGAON)\r\nNEAR V

RIVER MAPUSA ON
RIVER SAL PAZORKHONI,\r\nCUNCOLIM(NEAR CULVE

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RIVER SINQ

AMLAKHADI AFTE

BHADAR D/S JETPUR VILL. AFTER C

GIDC WATER RESERVOIR VILL:

TAPPAL
TRIVENI SANGAM, NR. SOMNATH

GHAGGAR BEFORE OTTU WEIR (BEFORE
ALHI RIVER-NOG(NEAR WATER SUPPLY\r\nSCHEME
HARABAGH(INTAKE FROM BBMB\r\nSOU
KUNNI PUL, VILL
MANJHI

MARKANDA

RAMPUR JATTAN MOGINAND NALA\r\nBEFORE CONFLUENCE TO RIVER\r\nMARKANDA
MA

HARMU RIVER B

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ON KHARKHAI RIVER

KUMARADHARA - U/S OF UPPINAGADY\r\nTOWN BEFORE C

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KYRHUKHLA NEA

GAUTAMI-GO
GAUTAM

CONFLUENCE POINT OF SUKHANA CHOE\r\nW

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953 RIVER NAGAVALI NEAR NH- 16 BRIDGE,\r\nUPSTREAM (U/S) OF SRIKAKULAM TOWN,
954 RIVER NAGAVALI NEAR WATER PUMP\r\nHOUSE (KILLIPALEM) , DOWN STREAM (D/S)\r\nOF SRIKAKULAM TOWN
955 RIVER VAMSADHARA, IMMEDIATE BORDER\r\nBETWEEN ANDHRA PRADESH & ORISSA\r\nSTATES, NEAR LALITHAMB
956 RIVER NAGAVALI IMMEDIATE BORDER\r\nBETWEEN ANDHRA PRADESH & ORISSA\r\nSTATES, NEAR SIT
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HAORA RIVER, D/S OF CHA

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RIVER KUJI (NEAR NEW BOGAGIGAON)\r\nNEAR V...

RIVER SAL PAZORKHONI, CUNCOLIM(NEAR\r\nCULVERT M...

RIVER MAPUSA ON...

RIVER SINQ...

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1026 RIVER AMLAKHADI AFTER
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1028 RIVER BHADAR D/S JETPUR VILL. AFTER\r\n## 1029
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1035 RIVER TRIVENI SANGAM, NR. SOMNATH
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1044 RIVER GHAGGAR BEFORE OTTU WEIR\r\n## 1045
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1052 RIVER MARKANDA
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1054 RIVER MANJHI
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1060 RIVER GIRI
1061 KUNNI PUL, VILLAGE
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1065 RIVER ALHI NOG(NEAR WATER SUPPLY\r\n## 1066 SCHEME
1067 RIVER HARABAGH(INTAKE FROM BBMB\r\n## 1068 SOUL
1069 RIVER GAJ WATER SUPPLY SCHEME\r\n## 1070 DHARAMSHALA
1071 RIVER BATHER WATER SUPPLY SCHEME

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RIVER KHARKHAD
RIVER HARMU B
RIVER KUMARADHARA - U/S OF\r\nUPPINAGADY TOWN B
BENNIOTHORA RESERVOIR
RIVI

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RIVER BANJAR NEAR ROAD BRIDGE

RIVER MAHANADI NEAR

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RIVER KYRHUKHLA NEAR

RIVER GAUT
RIVER

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RIVER CONFLUENCE POINT OF SUKHANA\

RIVER HAORA D/S OF CHA

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1344 RIVER NAGAVALI IMMEDIATE BORDER BETWEEN ANDHRA\r\nnPRADESH & ORISSA STATES, NEAR SIVAGANGAI
1345 RIVER NAGAVALI NEAR NH- 16 BRIDGE, UPSTREAM (U/S) OF\r\nnSRIKAKULAM TOWN,
1346 RIVER NAGAVALI NEAR WATER PUMP HOUSE (KILLIPALEM) ,\r\nnDOWN STREAM (D/S) OF SRIKAKULAM TOWN
1347 RIVER VAMSADHARA, IMMEDIATE BORDER BETWEEN\r\nnANDHRA PRADESH & ORISSA STATES, NEAR LALITHAMBUR
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1373 RIVER MAPUSA
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1377 RIVER SAL PAZOKHONI, CUNCOLIM(NEAR CULVERT\r\nn
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1387 RIVER AMLAKHADI AFTER CULVERT
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1390 RIVER BHADAR D/S JETPUR VILL. AFTER CULVERT
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1407 RIVER TRIVENI SANGAM, NR. SOMNATH
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1413 RIVER GHAGGAR BEFORE OTTU WEIR (BEF)
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1419 RIVER MARKAN
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1448 RIVER KUMARADHARA - U/S OF UPPINAGADY TOWN B
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1589 RIVER KYRHUKHLA
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1594 RIVER TLAWNG RIVER, SUARPUT, BA
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1611 RIVER TU
RIVER TUIRIAL
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RIVER G...

RIVER HAORA D/S OF CH...

	State	TemperatureMin	TemperatureMax
## 1	ANDHRA PRADESH	26.0	31.0
## 2	ANDHRA PRADESH	26.0	29.0
## 3	DAMAN AND DIU, DADRA\r\nAND NAGAR HAVELI	27.0	30.0
## 4	DAMAN AND DIU, DADRA\r\nAND NAGAR HAVELI	26.0	30.0
## 5	DAMAN AND DIU, DADRA\r\nAND NAGAR HAVELI	26.0	30.0
## 6	DAMAN AND DIU, DADRA\r\nAND NAGAR HAVELI	26.0	30.0
## 7	DAMAN AND DIU, DADRA\r\nAND NAGAR HAVELI	27.0	30.0
## 8	DAMAN AND DIU, DADRA\r\nAND NAGAR HAVELI	25.0	30.0
## 9	DAMAN AND DIU, DADRA\r\nAND NAGAR HAVELI	27.0	30.0
## 10	DAMAN AND DIU, DADRA\r\nAND NAGAR HAVELI	25.0	30.0
## 11	DAMAN AND DIU, DADRA\r\nAND NAGAR HAVELI	26.0	30.0
## 12	GOA	27.0	34.0
## 13	GOA	26.0	32.0
## 14	GOA	29.0	34.0
## 15	GOA	25.0	34.0
## 16	GOA	26.0	30.0
## 17	GOA	27.0	32.0
## 18	GOA	27.0	31.5
## 19	GOA	27.6	34.0
## 20	GOA	27.0	34.0
## 21	GOA	27.9	33.0
## 22	GOA	28.0	34.0
## 23	GOA	28.0	32.0
## 24	GOA	25.0	34.0
## 25	GOA	24.2	34.0
## 26	GOA	26.0	31.0
## 27	GOA	26.3	30.5
## 28	GOA	26.1	33.0
## 29	GOA	29.0	34.5
## 30	GOA	26.7	33.0
## 31	GOA	24.3	32.0
## 32	GOA	26.2	35.0
## 33	GOA	28.0	32.0
## 34	GOA	28.0	33.0
## 35	GUJARAT	24.0	33.0
## 36	GUJARAT	23.0	30.0
## 37	GUJARAT	23.0	30.0
## 38	GUJARAT	23.0	30.0
## 39	GUJARAT	28.0	32.0
## 40	GUJARAT	23.0	30.0
## 41	GUJARAT	27.0	31.0
## 42	GUJARAT	27.0	33.0
## 43	GUJARAT	27.0	34.0
## 44	GUJARAT	24.0	33.0
## 45	GUJARAT	24.0	33.0
## 46	GUJARAT	24.0	30.0
## 47	GUJARAT	28.0	35.0
## 48	GUJARAT	23.0	29.0
## 49	GUJARAT	22.0	25.0
## 50	GUJARAT	25.0	34.0
## 51	GUJARAT	23.0	30.0
## 52	HARYANA	<NA>	<NA>
## 53	HARYANA	<NA>	<NA>

## 54	HARYANA	<NA>	<NA>
## 55	HARYANA	<NA>	<NA>
## 56	HARYANA	<NA>	<NA>
## 57	HIMACHAL PRADESH	13.0	21.0
## 58	HIMACHAL PRADESH	15.0	25.0
## 59	HIMACHAL PRADESH	15.0	25.0
## 60	HIMACHAL PRADESH	13.0	23.0
## 61	HIMACHAL PRADESH	15.0	24.0
## 62	HIMACHAL PRADESH	13.0	24.0
## 63	KARNATAKA	26.0	28.0
## 64	KARNATAKA	27.0	33.0
## 65	KARNATAKA	28.0	33.0
## 66	KARNATAKA	25.0	28.0
## 67	KARNATAKA	28.0	33.0
## 68	KARNATAKA	28.0	33.0
## 69	KERALA	26.0	32.5
## 70	KERALA	26.0	29.7
## 71	KERALA	22.0	32.0
## 72	KERALA	25.0	32.0
## 73	KERALA	24.0	25.0
## 74	KERALA	27.0	30.0
## 75	KERALA	26.0	30.0
## 76	KERALA	30.0	33.0
## 77	KERALA	26.0	30.0
## 78	KERALA	26.0	30.1
## 79	KERALA	27.0	29.0
## 80	KERALA	27.0	30.0
## 81	KERALA	27.0	29.0
## 82	KERALA	25.0	28.0
## 83	KERALA	22.5	29.0
## 84	KERALA	30.0	33.0
## 85	KERALA	24.0	28.0
## 86	KERALA	29.0	33.0
## 87	KERALA	24.0	26.0
## 88	KERALA	27.0	29.0
## 89	KERALA	29.0	31.5
## 90	KERALA	26.0	32.0
## 91	KERALA	25.0	30.0
## 92	KERALA	25.0	32.0
## 93	KERALA	25.0	28.0
## 94	KERALA	24.0	28.0
## 95	KERALA	25.0	29.0
## 96	KERALA	26.0	39.0
## 97	KERALA	30.0	33.0
## 98	KERALA	30.0	33.0
## 99	KERALA	30.0	33.0
## 100	KERALA	21.0	25.0
## 101	KERALA	25.0	30.0
## 102	KERALA	27.0	32.0
## 103	KERALA	26.0	31.0
## 104	KERALA	26.0	32.0
## 105	KERALA	29.5	31.5
## 106	KERALA	29.0	31.5
## 107	KERALA	22.0	30.0

## 108	KERALA	24.0	31.0
## 109	KERALA	25.0	28.0
## 110	KERALA	24.0	29.0
## 111	KERALA	23.0	32.0
## 112	KERALA	26.0	31.0
## 113	KERALA	24.0	30.0
## 114	KERALA	24.0	30.0
## 115	KERALA	24.0	29.0
## 116	KERALA	24.0	29.0
## 117	KERALA	24.0	32.0
## 118	KERALA	24.0	29.0
## 119	KERALA	25.0	32.0
## 120	KERALA	24.0	30.0
## 121	KERALA	25.0	30.0
## 122	KERALA	28.0	31.0
## 123	KERALA	26.0	31.0
## 124	KERALA	26.0	30.0
## 125	KERALA	22.2	32.0
## 126	KERALA	26.0	32.0
## 127	KERALA	26.0	32.0
## 128	KERALA	26.0	31.0
## 129	KERALA	26.8	33.0
## 130	KERALA	26.0	32.0
## 131	KERALA	27.1	28.2
## 132	KERALA	26.8	28.2
## 133	KERALA	28.0	30.0
## 134	KERALA	24.0	32.0
## 135	KERALA	24.0	28.0
## 136	KERALA	25.0	28.0
## 137	KERALA	25.0	32.0
## 138	KERALA	27.0	33.0
## 139	KERALA	25.0	32.0
## 140	MAHARASHTRA	27.0	35.0
## 141	MAHARASHTRA	27.0	33.0
## 142	MAHARASHTRA	26.0	33.0
## 143	MAHARASHTRA	20.0	28.0
## 144	MAHARASHTRA	25.0	28.0
## 145	MAHARASHTRA	20.0	32.0
## 146	MAHARASHTRA	20.0	28.0
## 147	MAHARASHTRA	27.0	33.0
## 148	MAHARASHTRA	22.0	24.0
## 149	MAHARASHTRA	23.0	33.0
## 150	MAHARASHTRA	25.0	28.0
## 151	MAHARASHTRA	27.0	36.0
## 152	MAHARASHTRA	25.0	28.0
## 153	MAHARASHTRA	20.0	36.0
## 154	MAHARASHTRA	21.0	38.0
## 155	MAHARASHTRA	22.0	24.0
## 156	MAHARASHTRA	25.0	28.0
## 157	MAHARASHTRA	25.0	28.0
## 158	MAHARASHTRA	21.0	37.0
## 159	MAHARASHTRA	20.0	28.0
## 160	MAHARASHTRA	20.0	28.0
## 161	MAHARASHTRA	20.0	28.0

## 162	MAHARASHTRA	20.0	28.0
## 163	MAHARASHTRA	20.0	28.0
## 164	MAHARASHTRA	27.0	34.0
## 165	MAHARASHTRA	26.0	36.0
## 166	MAHARASHTRA	26.0	35.0
## 167	MAHARASHTRA	26.0	36.0
## 168	MAHARASHTRA	27.0	36.0
## 169	MAHARASHTRA	25.0	30.0
## 170	MAHARASHTRA	25.0	30.0
## 171	MAHARASHTRA	25.0	32.0
## 172	MAHARASHTRA	23.0	26.0
## 173	MAHARASHTRA	24.0	30.0
## 174	MAHARASHTRA	22.0	24.0
## 175	MAHARASHTRA	22.0	24.0
## 176	MANIPUR	22.1	24.3
## 177	MANIPUR	21.0	24.1
## 178	MANIPUR	21.7	24.0
## 179	MANIPUR	21.3	24.0
## 180	MANIPUR	22.1	23.8
## 181	MANIPUR	22.4	23.2
## 182	MANIPUR	23.0	24.0
## 183	MANIPUR	22.3	22.6
## 184	MANIPUR	20.3	25.0
## 185	MANIPUR	21.4	24.2
## 186	MANIPUR	21.6	24.8
## 187	MANIPUR	21.0	22.0
## 188	MANIPUR	21.4	23.2
## 189	MANIPUR	21.0	23.1
## 190	MANIPUR	20.6	22.9
## 191	MANIPUR	21.8	22.7
## 192	MANIPUR	22.2	24.0
## 193	MANIPUR	21.4	23.1
## 194	MANIPUR	22.1	23.9
## 195	MANIPUR	22.3	23.8
## 196	MANIPUR	21.0	22.5
## 197	MANIPUR	21.0	22.5
## 198	MANIPUR	20.8	23.7
## 199	MANIPUR	20.9	23.1
## 200	MANIPUR	20.1	23.1
## 201	MANIPUR	20.0	22.6
## 202	MANIPUR	20.1	22.7
## 203	MANIPUR	19.0	21.6
## 204	MANIPUR	21.4	23.0
## 205	MANIPUR	20.0	22.5
## 206	MANIPUR	21.6	24.0
## 207	MANIPUR	21.0	22.6
## 208	MANIPUR	21.6	24.0
## 209	MEGHALAYA	17.0	25.0
## 210	MEGHALAYA	15.0	24.0
## 211	MEGHALAYA	15.0	22.0
## 212	MEGHALAYA	16.0	28.0
## 213	MEGHALAYA	17.0	28.0
## 214	MEGHALAYA	13.0	21.0
## 215	MEGHALAYA	15.0	24.0

## 216	MEGHALAYA	14.0	23.0
## 217	MEGHALAYA	13.0	22.0
## 218	MEGHALAYA	12.0	20.0
## 219	MEGHALAYA	15.0	21.0
## 220	MEGHALAYA	15.0	28.0
## 221	MEGHALAYA	13.0	21.0
## 222	MEGHALAYA	15.0	26.0
## 223	MEGHALAYA	13.0	22.0
## 224	MEGHALAYA	13.0	21.0
## 225	MEGHALAYA	13.0	24.0
## 226	MEGHALAYA	13.0	23.0
## 227	MEGHALAYA	10.0	21.0
## 228	MEGHALAYA	13.0	22.0
## 229	MEGHALAYA	13.0	23.5
## 230	MEGHALAYA	16.0	27.0
## 231	MEGHALAYA	15.0	25.0
## 232	MEGHALAYA	15.0	26.0
## 233	MEGHALAYA	14.0	23.0
## 234	MEGHALAYA	19.0	28.0
## 235	MEGHALAYA	20.0	32.0
## 236	PUDUCHERRY	30.0	30.0
## 237	PUDUCHERRY	30.0	30.0
## 238	PUDUCHERRY	30.0	30.0
## 239	PUDUCHERRY	28.0	28.0
## 240	PUNJAB	17.0	32.0
## 241	PUNJAB	9.0	28.0
## 242	PUNJAB	16.0	32.0
## 243	PUNJAB	8.0	29.0
## 244	PUNJAB	19.0	34.0
## 245	PUNJAB	16.0	31.0
## 246	PUNJAB	15.0	32.0
## 247	PUNJAB	16.0	33.0
## 248	PUNJAB	15.0	32.0
## 249	PUNJAB	9.0	28.5
## 250	PUNJAB	19.0	34.0
## 251	PUNJAB	18.0	33.0
## 252	PUNJAB	17.0	35.0
## 253	RAJASTHAN	13.0	30.0
## 254	RAJASTHAN	14.0	28.0
## 255	RAJASTHAN	12.0	29.0
## 256	TAMIL NADU	21.0	29.0
## 257	TAMIL NADU	22.0	31.0
## 258	TAMIL NADU	23.0	30.0
## 259	TAMIL NADU	29.0	29.0
## 260	TAMIL NADU	22.0	31.0
## 261	TAMIL NADU	21.0	29.0
## 262	TAMIL NADU	29.0	29.0
## 263	TAMIL NADU	29.0	29.0
## 264	TAMIL NADU	21.0	30.0
## 265	TAMIL NADU	29.0	29.0
## 266	TAMIL NADU	23.0	29.0
## 267	TAMIL NADU	24.0	29.0
## 268	TRIPURA	24.5	27.0
## 269	TRIPURA	23.5	27.0

## 270	TRIPURA	25.0	27.0
## 271	TRIPURA	25.0	27.0
## 272	TRIPURA	24.5	27.0
## 273	TRIPURA	24.5	27.0
## 274	TRIPURA	25.0	27.0
## 275	TRIPURA	25.0	27.0
## 276	TRIPURA	25.0	27.0
## 277	TRIPURA	25.0	27.0
## 278	TRIPURA	25.0	27.0
## 279	TRIPURA	25.0	27.0
## 280	TRIPURA	25.0	27.0
## 281	TRIPURA	25.0	27.0
## 282	TRIPURA	24.0	27.0
## 283	TRIPURA	25.0	27.0
## 284	TRIPURA	25.0	27.0
## 285	TRIPURA	25.0	27.0
## 286	TRIPURA	25.0	27.0
## 287	TRIPURA	25.0	27.0
## 288	TRIPURA	24.5	27.0
## 289	TRIPURA	24.5	27.0
## 290	TRIPURA	24.5	27.0
## 291	TRIPURA	25.0	27.0
## 292	TRIPURA	25.0	27.0
## 293	TRIPURA	25.0	27.0
## 294	ANDHRA PRADESH	25.0	30.0
## 295	ANDHRA PRADESH	25.0	30.0
## 296	DAMAN, DIU, DADRA\r\nNAGAR HAVELI	27.6	32.1
## 297	DAMAN, DIU, DADRA\r\nNAGAR HAVELI	28.0	31.4
## 298	DAMAN, DIU, DADRA\r\nNAGAR HAVELI	28.0	31.2
## 299	DAMAN, DIU, DADRA\r\nNAGAR HAVELI	28.0	32.1
## 300	DAMAN, DIU, DADRA\r\nNAGAR HAVELI	28.0	32.0
## 301	DAMAN, DIU, DADRA\r\nNAGAR HAVELI	28.0	33.0
## 302	DAMAN, DIU, DADRA\r\nNAGAR HAVELI	28.0	32.2
## 303	DAMAN, DIU, DADRA\r\nNAGAR HAVELI	28.0	32.0
## 304	DAMAN, DIU, DADRA\r\nNAGAR HAVELI	29.0	31.8
## 305	GOA	26.8	31.2
## 306	GOA	26.6	34.0
## 307	GOA	26.0	33.0
## 308	GOA	26.0	31.2
## 309	GOA	25.5	32.0
## 310	GOA	26.0	32.2
## 311	GOA	25.0	32.2
## 312	GOA	27.0	32.0
## 313	GOA	26.0	31.0
## 314	GOA	25.2	32.0
## 315	GOA	26.0	32.0
## 316	GOA	26.0	30.0
## 317	GOA	27.0	31.1
## 318	GOA	27.0	32.5
## 319	GOA	26.1	34.0
## 320	GOA	26.0	33.0
## 321	GOA	26.0	32.0
## 322	GOA	26.0	34.0
## 323	GOA	26.0	33.0

## 324	GOA	25.2	30.0
## 325	GOA	24.9	33.0
## 326	GOA	27.0	31.1
## 327	GOA	27.0	31.5
## 328	GOA	26.9	34.0
## 329	GUJARAT	26.0	34.0
## 330	GUJARAT	25.0	31.0
## 331	GUJARAT	20.0	31.0
## 332	GUJARAT	23.0	31.0
## 333	GUJARAT	28.0	33.0
## 334	GUJARAT	25.0	31.0
## 335	GUJARAT	28.0	31.0
## 336	GUJARAT	<NA>	35.0
## 337	GUJARAT	<NA>	36.0
## 338	GUJARAT	26.0	33.0
## 339	GUJARAT	26.0	34.0
## 340	GUJARAT	<NA>	30.0
## 341	GUJARAT	29.0	32.0
## 342	GUJARAT	25.0	35.0
## 343	GUJARAT	20.0	24.0
## 344	GUJARAT	<NA>	32.0
## 345	GUJARAT	<NA>	31.0
## 346	GUJARAT	<NA>	<NA>
## 347	GUJARAT	23.0	28.0
## 348	HARYANA	<NA>	<NA>
## 349	HARYANA	<NA>	<NA>
## 350	HARYANA	<NA>	<NA>
## 351	HARYANA	<NA>	<NA>
## 352	HARYANA	<NA>	<NA>
## 353	KARNATAKA	26.0	31.7
## 354	KARNATAKA	25.0	29.0
## 355	KARNATAKA	26.0	31.0
## 356	KARNATAKA	26.0	34.0
## 357	KERALA	24.0	32.0
## 358	KERALA	26.0	31.0
## 359	KERALA	24.0	33.0
## 360	KERALA	24.0	32.0
## 361	KERALA	23.0	29.0
## 362	KERALA	26.0	29.0
## 363	KERALA	26.0	32.0
## 364	KERALA	31.0	32.0
## 365	KERALA	27.0	30.0
## 366	KERALA	25.0	32.0
## 367	KERALA	25.0	28.0
## 368	KERALA	26.0	32.0
## 369	KERALA	27.0	30.0
## 370	KERALA	21.0	28.0
## 371	KERALA	26.0	32.0
## 372	KERALA	30.0	32.0
## 373	KERALA	23.0	28.0
## 374	KERALA	31.0	32.0
## 375	KERALA	22.0	28.0
## 376	KERALA	27.0	30.0
## 377	KERALA	28.0	31.0

## 378	KERALA	27.0	33.0
## 379	KERALA	25.0	30.0
## 380	KERALA	24.0	31.0
## 381	KERALA	27.0	32.0
## 382	KERALA	28.0	32.0
## 383	KERALA	28.0	32.0
## 384	KERALA	26.0	30.0
## 385	KERALA	30.0	32.0
## 386	KERALA	30.0	32.0
## 387	KERALA	31.0	32.0
## 388	KERALA	20.0	30.0
## 389	KERALA	23.0	30.0
## 390	KERALA	27.0	32.0
## 391	KERALA	26.0	31.0
## 392	KERALA	29.0	31.0
## 393	KERALA	28.0	31.5
## 394	KERALA	28.0	31.5
## 395	KERALA	28.0	32.0
## 396	KERALA	26.0	31.0
## 397	KERALA	25.0	29.0
## 398	KERALA	24.0	32.0
## 399	KERALA	24.0	30.0
## 400	KERALA	25.0	32.0
## 401	KERALA	26.0	31.0
## 402	KERALA	25.0	30.0
## 403	KERALA	28.0	32.0
## 404	KERALA	28.0	34.0
## 405	KERALA	26.0	32.0
## 406	KERALA	28.0	31.0
## 407	KERALA	27.0	34.0
## 408	KERALA	27.0	32.0
## 409	KERALA	24.0	32.0
## 410	KERALA	24.0	33.0
## 411	KERALA	27.0	31.0
## 412	KERALA	27.0	32.0
## 413	KERALA	25.0	33.0
## 414	KERALA	25.0	30.0
## 415	KERALA	25.0	31.0
## 416	KERALA	25.0	30.0
## 417	KERALA	26.0	30.0
## 418	KERALA	27.0	30.0
## 419	KERALA	25.0	31.0
## 420	KERALA	25.0	32.0
## 421	KERALA	25.0	32.0
## 422	KERALA	26.0	32.0
## 423	KERALA	20.0	30.0
## 424	KERALA	22.0	30.0
## 425	KERALA	25.0	32.0
## 426	KERALA	31.0	34.0
## 427	KERALA	24.0	32.0
## 428	KERALA	27.0	31.0
## 429	KERALA	27.0	31.0
## 430	MAHARASHTRA	28.0	28.0
## 431	MAHARASHTRA	27.0	27.0

## 432	MANIPUR	18.0	30.0
## 433	MANIPUR	17.0	29.0
## 434	MANIPUR	17.0	30.0
## 435	MANIPUR	19.0	28.0
## 436	MANIPUR	19.0	30.0
## 437	MANIPUR	19.0	30.0
## 438	MANIPUR	18.0	30.0
## 439	MANIPUR	17.0	29.0
## 440	MANIPUR	18.0	18.0
## 441	MANIPUR	18.0	18.0
## 442	MANIPUR	17.0	17.0
## 443	MANIPUR	16.3	29.1
## 444	MANIPUR	17.0	29.0
## 445	MANIPUR	17.0	28.0
## 446	MANIPUR	19.0	28.0
## 447	MANIPUR	18.0	29.0
## 448	MANIPUR	6.9	30.0
## 449	MANIPUR	17.3	29.0
## 450	MANIPUR	18.0	30.0
## 451	MANIPUR	19.0	29.4
## 452	MANIPUR	15.0	30.0
## 453	MANIPUR	7.3	28.3
## 454	MANIPUR	17.0	29.0
## 455	MANIPUR	19.2	19.2
## 456	MANIPUR	12.8	28.7
## 457	MANIPUR	18.0	29.0
## 458	MANIPUR	17.0	29.0
## 459	MANIPUR	18.0	30.7
## 460	MANIPUR	29.0	29.0
## 461	MANIPUR	18.0	29.7
## 462	MANIPUR	15.6	29.9
## 463	MEGHALAYA	16.0	26.0
## 464	MEGHALAYA	15.0	25.0
## 465	MEGHALAYA	10.0	22.0
## 466	MEGHALAYA	14.0	28.0
## 467	MEGHALAYA	15.0	31.0
## 468	MEGHALAYA	7.2	19.0
## 469	MEGHALAYA	6.9	25.0
## 470	MEGHALAYA	13.0	23.0
## 471	MEGHALAYA	10.0	21.0
## 472	MEGHALAYA	10.0	23.0
## 473	MEGHALAYA	9.0	21.0
## 474	MEGHALAYA	15.0	28.0
## 475	MEGHALAYA	9.0	21.0
## 476	MEGHALAYA	14.0	28.0
## 477	MEGHALAYA	10.0	22.0
## 478	MEGHALAYA	11.0	21.0
## 479	MEGHALAYA	9.0	22.0
## 480	MEGHALAYA	9.0	22.0
## 481	MEGHALAYA	8.0	20.0
## 482	MEGHALAYA	9.0	21.0
## 483	MEGHALAYA	10.0	22.0
## 484	MEGHALAYA	14.0	24.0
## 485	MEGHALAYA	15.0	26.0

## 486	MEGHALAYA	16.0	26.0
## 487	MEGHALAYA	15.0	24.0
## 488	MEGHALAYA	18.0	30.0
## 489	MEGHALAYA	18.0	30.0
## 490	ODISHA	10.0	26.0
## 491	ODISHA	10.0	26.0
## 492	ODISHA	18.0	31.0
## 493	ODISHA	16.0	28.0
## 494	ODISHA	18.0	31.0
## 495	ODISHA	14.0	31.0
## 496	ODISHA	16.0	28.0
## 497	ODISHA	12.0	26.0
## 498	ODISHA	19.0	32.0
## 499	ODISHA	15.0	33.0
## 500	PUDUCHERRY	29.0	31.0
## 501	PUDUCHERRY	31.0	31.0
## 502	PUDUCHERRY	31.0	31.0
## 503	PUDUCHERRY	31.0	31.0
## 504	PUDUCHERRY	30.0	30.0
## 505	PUNJAB	14.0	29.0
## 506	PUNJAB	14.0	30.0
## 507	PUNJAB	15.0	28.0
## 508	PUNJAB	11.0	29.0
## 509	PUNJAB	15.0	28.0
## 510	PUNJAB	12.0	27.0
## 511	PUNJAB	12.0	28.0
## 512	PUNJAB	14.0	29.0
## 513	PUNJAB	15.0	29.0
## 514	PUNJAB	15.0	31.0
## 515	PUNJAB	14.0	30.0
## 516	PUNJAB	13.0	29.0
## 517	PUNJAB	20.0	33.0
## 518	RAJASTHAN	10.0	27.0
## 519	RAJASTHAN	25.3	25.3
## 520	RAJASTHAN	26.6	26.6
## 521	RAJASTHAN	12.0	29.0
## 522	RAJASTHAN	13.0	18.0
## 523	RAJASTHAN	11.0	28.5
## 524	RAJASTHAN	13.5	28.5
## 525	TAMIL NADU	22.0	24.0
## 526	TAMIL NADU	23.0	27.0
## 527	TAMIL NADU	24.0	28.0
## 528	TAMIL NADU	29.0	29.0
## 529	TAMIL NADU	23.0	25.0
## 530	TAMIL NADU	23.0	24.0
## 531	TAMIL NADU	6.9	29.0
## 532	TAMIL NADU	24.0	24.0
## 533	TAMIL NADU	29.0	29.0
## 534	TAMIL NADU	23.0	25.0
## 535	TAMIL NADU	29.0	29.0
## 536	TAMIL NADU	24.0	28.0
## 537	TAMIL NADU	25.0	28.0
## 538	TRIPURA	20.0	29.5
## 539	TRIPURA	20.5	29.0

## 540	TRIPURA	20.5	31.0
## 541	TRIPURA	20.0	29.5
## 542	TRIPURA	20.5	29.0
## 543	TRIPURA	20.5	29.0
## 544	TRIPURA	20.0	29.0
## 545	TRIPURA	20.5	29.5
## 546	TRIPURA	20.5	29.0
## 547	TRIPURA	20.5	28.5
## 548	TRIPURA	20.5	29.0
## 549	TRIPURA	20.0	29.5
## 550	TRIPURA	20.5	29.0
## 551	TRIPURA	20.5	29.0
## 552	TRIPURA	20.5	28.5
## 553	TRIPURA	20.5	29.0
## 554	TRIPURA	20.0	28.5
## 555	TRIPURA	20.0	29.0
## 556	TRIPURA	20.5	29.5
## 557	TRIPURA	20.5	29.0
## 558	TRIPURA	20.0	28.5
## 559	TRIPURA	20.5	28.5
## 560	TRIPURA	20.0	28.0
## 561	TRIPURA	20.0	28.5
## 562	TRIPURA	20.0	28.5
## 563	TRIPURA	20.0	28.5
## 564	ANDHRA PRADESH	25.0	31.0
## 565	ANDHRA PRADESH	24.0	36.0
## 566	ANDHRA PRADESH	21.0	30.0
## 567	ANDHRA PRADESH	21.0	30.0
## 568	ANDHRA PRADESH	20.0	34.0
## 569	ANDHRA PRADESH	24.0	34.0
## 570	ANDHRA PRADESH	24.0	30.0
## 571	ANDHRA PRADESH	25.0	32.0
## 572	ANDHRA PRADESH	25.0	28.0
## 573	ANDHRA PRADESH	20.0	33.0
## 574	ANDHRA PRADESH	25.0	26.0
## 575	ASSAM	20.0	30.0
## 576	ASSAM	21.0	32.0
## 577	ASSAM	19.0	26.0
## 578	ASSAM	12.0	31.0
## 579	ASSAM	21.0	34.0
## 580	ASSAM	18.0	33.0
## 581	ASSAM	23.0	36.0
## 582	ASSAM	19.0	31.0
## 583	ASSAM	18.0	30.0
## 584	ASSAM	13.0	32.0
## 585	ASSAM	13.0	32.0
## 586	ASSAM	18.0	33.0
## 587	ASSAM	20.0	30.0
## 588	ASSAM	20.0	34.0
## 589	ASSAM	19.0	30.0
## 590	ASSAM	24.0	34.0
## 591	ASSAM	13.0	32.0
## 592	ASSAM	22.0	25.0
## 593	ASSAM	12.0	31.0

## 594	ASSAM	22.0	25.0
## 595	ASSAM	13.0	25.0
## 596	ASSAM	20.0	32.0
## 597	ASSAM	21.0	40.0
## 598	ASSAM	20.0	32.0
## 599	ASSAM	20.0	33.0
## 600	ASSAM	14.0	31.0
## 601	ASSAM	23.0	34.0
## 602	ASSAM	20.0	34.0
## 603	ASSAM	18.0	29.0
## 604	ASSAM	20.0	31.0
## 605	ASSAM	13.0	30.0
## 606	ASSAM	19.0	31.0
## 607	BIHAR	25.0	36.0
## 608	BIHAR	18.0	33.0
## 609	BIHAR	16.0	30.0
## 610	BIHAR	18.0	31.0
## 611	HARYANA	-	-
## 612	HARYANA	-	-
## 613	HARYANA	-	-
## 614	PUNJAB	-	-
## 615	HIMACHAL\r\nPRADESH	-	-
## 616	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	28.0	33.0
## 617	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	28.0	34.0
## 618	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	29.0	34.0
## 619	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	28.0	33.0
## 620	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	27.5	34.0
## 621	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	27.5	33.0
## 622	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	28.0	34.0
## 623	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	27.5	33.0
## 624	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	28.0	33.0
## 625	GOA	27.0	31.0
## 626	GOA	27.0	31.0
## 627	GOA	27.0	30.0
## 628	GOA	27.0	30.0
## 629	GOA	27.0	30.0
## 630	GOA	27.0	31.0
## 631	GOA	27.0	31.0
## 632	GOA	27.0	37.0
## 633	GOA	27.0	31.0
## 634	GOA	27.0	31.0
## 635	GOA	27.0	30.0
## 636	GOA	27.0	31.0
## 637	GOA	27.0	31.0
## 638	GOA	27.0	30.0
## 639	GOA	28.0	31.0
## 640	GOA	28.0	30.0
## 641	GOA	27.0	31.0
## 642	GOA	27.0	29.0
## 643	GOA	27.0	33.0
## 644	GOA	26.0	32.0
## 645	GOA	27.0	32.0
## 646	GOA	27.0	31.0
## 647	GOA	27.0	32.0

## 648	GOA	27.0	31.0
## 649	GOA	27.0	31.0
## 650	GOA	27.0	31.0
## 651	GUJARAT	25.0	35.0
## 652	GUJARAT	28.0	34.0
## 653	GUJARAT	28.0	33.0
## 654	GUJARAT	27.0	31.0
## 655	GUJARAT	25.0	31.0
## 656	GUJARAT	22.0	29.0
## 657	GUJARAT	22.0	33.0
## 658	GUJARAT	27.0	32.0
## 659	GUJARAT	27.0	32.0
## 660	GUJARAT	24.0	35.0
## 661	GUJARAT	24.0	32.0
## 662	GUJARAT	26.0	31.0
## 663	GUJARAT	24.0	31.0
## 664	GUJARAT	24.0	35.0
## 665	GUJARAT	30.0	30.0
## 666	GUJARAT	22.0	26.0
## 667	GUJARAT	20.0	29.0
## 668	GUJARAT	27.0	30.0
## 669	HARYANA	18.0	18.0
## 670	HARYANA	-	-
## 671	HARYANA	18.0	29.0
## 672	HARYANA	16.0	29.0
## 673	HARYANA	-	-
## 674	HIMACHAL\r\nPRADESH	19.0	25.0
## 675	HIMACHAL\r\nPRADESH	20.0	22.0
## 676	HIMACHAL\r\nPRADESH	20.0	20.6
## 677	HIMACHAL\r\nPRADESH	8.0	11.0
## 678	HIMACHAL\r\nPRADESH	9.0	12.0
## 679	HIMACHAL\r\nPRADESH	11.0	29.0
## 680	HIMACHAL\r\nPRADESH	12.0	32.0
## 681	HIMACHAL\r\nPRADESH	13.0	32.0
## 682	HIMACHAL\r\nPRADESH	12.0	32.0
## 683	HIMACHAL\r\nPRADESH	12.0	29.0
## 684	HIMACHAL\r\nPRADESH	13.0	33.0
## 685	HIMACHAL\r\nPRADESH	11.0	13.5
## 686	HIMACHAL\r\nPRADESH	11.5	13.5
## 687	HIMACHAL\r\nPRADESH	5.0	18.0
## 688	JAMMU &\r\nKASHMIR	22.0	22.0
## 689	JAMMU &\r\nKASHMIR	21.0	22.0
## 690	JAMMU &\r\nKASHMIR	16.0	16.0
## 691	JHARKHAND	19.5	33.0
## 692	JHARKHAND	8.0	27.0
## 693	JHARKHAND	9.0	28.0
## 694	JHARKHAND	16.0	35.5
## 695	JHARKHAND	16.0	38.0
## 696	JHARKHAND	15.0	27.0
## 697	JHARKHAND	19.0	35.0
## 698	JHARKHAND	12.0	25.0
## 699	KARNATAKA	27.0	30.0
## 700	KARNATAKA	26.0	29.0
## 701	KARNATAKA	23.0	29.0

## 702	KARNATAKA	23.0	26.0
## 703	KARNATAKA	25.0	27.0
## 704	KARNATAKA	25.0	31.0
## 705	KARNATAKA	25.0	32.0
## 706	KARNATAKA	26.0	30.0
## 707	KARNATAKA	25.0	30.0
## 708	KERALA	19.0	24.0
## 709	KERALA	19.0	24.0
## 710	KERALA	20.0	23.0
## 711	KERALA	20.0	23.0
## 712	KERALA	23.0	33.0
## 713	KERALA	22.0	32.0
## 714	KERALA	27.0	33.0
## 715	KERALA	22.0	32.0
## 716	KERALA	22.0	33.0
## 717	KERALA	26.0	31.0
## 718	KERALA	22.0	31.0
## 719	KERALA	23.0	31.0
## 720	KERALA	20.0	32.0
## 721	KERALA	29.0	34.0
## 722	KERALA	25.0	30.0
## 723	KERALA	26.0	32.0
## 724	KERALA	18.0	23.0
## 725	KERALA	19.0	24.0
## 726	KERALA	22.0	33.0
## 727	KERALA	22.0	33.0
## 728	KERALA	26.0	31.0
## 729	KERALA	26.0	31.0
## 730	KERALA	18.0	25.0
## 731	KERALA	23.0	34.0
## 732	KERALA	24.0	31.0
## 733	KERALA	27.0	32.0
## 734	KERALA	25.0	32.0
## 735	KERALA	28.0	37.0
## 736	KERALA	28.0	37.0
## 737	KERALA	26.0	32.0
## 738	KERALA	22.0	32.0
## 739	KERALA	23.0	32.0
## 740	KERALA	25.0	37.0
## 741	KERALA	24.0	31.0
## 742	KERALA	21.0	32.0
## 743	KERALA	26.0	32.0
## 744	KERALA	24.0	30.0
## 745	KERALA	28.0	34.0
## 746	KERALA	26.0	32.0
## 747	KERALA	26.0	32.0
## 748	KERALA	26.0	34.0
## 749	KERALA	27.0	30.0
## 750	KERALA	27.0	31.0
## 751	KERALA	28.0	32.0
## 752	KERALA	28.0	32.0
## 753	KERALA	26.0	32.0
## 754	KERALA	26.0	34.0
## 755	KERALA	28.0	35.0

## 756	KERALA	28.0	34.0
## 757	KERALA	18.0	24.0
## 758	KERALA	28.0	32.0
## 759	KERALA	27.0	32.0
## 760	KERALA	24.0	29.0
## 761	KERALA	25.0	30.0
## 762	KERALA	27.0	30.0
## 763	KERALA	27.0	32.0
## 764	KERALA	25.0	30.0
## 765	KERALA	24.0	31.0
## 766	KERALA	24.0	33.0
## 767	KERALA	24.0	31.0
## 768	KERALA	25.0	31.0
## 769	KERALA	26.0	31.0
## 770	KERALA	23.0	31.0
## 771	KERALA	22.0	32.0
## 772	KERALA	28.0	30.0
## 773	KERALA	22.0	32.0
## 774	KERALA	21.0	34.0
## 775	KERALA	28.0	31.0
## 776	KERALA	24.0	32.0
## 777	KERALA	22.0	34.0
## 778	KERALA	28.0	32.0
## 779	KERALA	24.0	32.0
## 780	KERALA	27.0	35.0
## 781	MADHYA PRADESH	19.0	28.0
## 782	MADHYA PRADESH	15.0	29.0
## 783	MADHYA PRADESH	22.0	34.0
## 784	MADHYA PRADESH	18.2	22.1
## 785	MADHYA PRADESH	24.5	27.6
## 786	MADHYA PRADESH	23.0	30.0
## 787	MADHYA PRADESH	21.0	28.0
## 788	MADHYA PRADESH	22.0	38.0
## 789	MADHYA PRADESH	26.5	33.0
## 790	MAHARASHTRA	25.0	38.0
## 791	MAHARASHTRA	20.0	30.0
## 792	MAHARASHTRA	16.0	39.0
## 793	MAHARASHTRA	15.0	38.0
## 794	MAHARASHTRA	24.0	30.0
## 795	MAHARASHTRA	26.0	32.0
## 796	MAHARASHTRA	25.0	28.0
## 797	MAHARASHTRA	20.0	28.0
## 798	MAHARASHTRA	20.0	28.0
## 799	MAHARASHTRA	25.0	36.0
## 800	MAHARASHTRA	22.0	37.0
## 801	MAHARASHTRA	22.0	30.0
## 802	MAHARASHTRA	28.0	29.0
## 803	MAHARASHTRA	18.0	29.0
## 804	MAHARASHTRA	18.0	29.0
## 805	MAHARASHTRA	18.0	29.0
## 806	MAHARASHTRA	28.0	29.0
## 807	MAHARASHTRA	28.0	29.0
## 808	MAHARASHTRA	26.0	29.0
## 809	MAHARASHTRA	15.0	40.0

## 810	MAHARASHTRA	17.0	35.0
## 811	MAHARASHTRA	17.0	35.0
## 812	MAHARASHTRA	17.0	35.0
## 813	MAHARASHTRA	17.0	35.0
## 814	MAHARASHTRA	17.0	35.0
## 815	MAHARASHTRA	25.0	30.0
## 816	MAHARASHTRA	25.0	31.0
## 817	MAHARASHTRA	25.0	31.0
## 818	MAHARASHTRA	21.0	32.0
## 819	MAHARASHTRA	25.0	32.0
## 820	MAHARASHTRA	24.0	32.0
## 821	MAHARASHTRA	25.0	32.0
## 822	MAHARASHTRA	22.0	32.0
## 823	MAHARASHTRA	26.0	28.0
## 824	MAHARASHTRA	26.0	28.0
## 825	MAHARASHTRA	26.0	28.0
## 826	MANIPUR	17.0	30.0
## 827	MANIPUR	16.0	29.0
## 828	MANIPUR	17.0	30.0
## 829	MANIPUR	17.0	28.0
## 830	MANIPUR	18.0	29.0
## 831	MANIPUR	17.0	30.0
## 832	MANIPUR	16.0	30.0
## 833	MANIPUR	19.0	19.0
## 834	MANIPUR	17.0	30.0
## 835	MANIPUR	18.0	30.0
## 836	MANIPUR	16.0	29.0
## 837	MANIPUR	17.0	28.0
## 838	MANIPUR	16.0	30.0
## 839	MANIPUR	17.0	30.0
## 840	MANIPUR	18.0	31.0
## 841	MANIPUR	16.0	29.0
## 842	MANIPUR	17.0	30.0
## 843	MANIPUR	15.0	28.0
## 844	MANIPUR	17.0	28.0
## 845	MANIPUR	15.0	28.0
## 846	MANIPUR	19.0	31.0
## 847	MANIPUR	19.0	31.0
## 848	MANIPUR	18.0	30.0
## 849	MANIPUR	18.0	31.0
## 850	MANIPUR	19.0	30.0
## 851	MANIPUR	17.0	28.2
## 852	MANIPUR	16.0	28.0
## 853	MANIPUR	17.0	28.3
## 854	MANIPUR	19.0	29.0
## 855	MANIPUR	19.0	30.0
## 856	MANIPUR	16.0	28.0
## 857	MANIPUR	17.0	29.0
## 858	MEGHALAYA	17.0	28.0
## 859	MEGHALAYA	17.0	30.0
## 860	MEGHALAYA	17.0	30.0
## 861	MEGHALAYA	18.0	30.9
## 862	MEGHALAYA	15.0	26.0
## 863	MEGHALAYA	14.0	26.0

## 864	MEGHALAYA	15.0	32.0
## 865	MEGHALAYA	12.0	22.0
## 866	MEGHALAYA	14.0	27.0
## 867	MEGHALAYA	18.0	29.0
## 868	MEGHALAYA	17.0	30.2
## 869	MEGHALAYA	18.0	30.0
## 870	MEGHALAYA	10.0	19.0
## 871	MEGHALAYA	9.0	22.0
## 872	MEGHALAYA	9.0	21.0
## 873	MEGHALAYA	16.0	24.0
## 874	MEGHALAYA	13.0	19.0
## 875	MEGHALAYA	13.0	19.0
## 876	MEGHALAYA	13.0	19.0
## 877	MEGHALAYA	9.0	20.0
## 878	MEGHALAYA	10.0	19.0
## 879	MEGHALAYA	9.0	19.0
## 880	MEGHALAYA	9.0	19.0
## 881	MEGHALAYA	10.0	19.0
## 882	MEGHALAYA	18.0	26.0
## 883	MEGHALAYA	9.0	20.0
## 884	MEGHALAYA	9.0	19.0
## 885	MEGHALAYA	9.0	18.0
## 886	MEGHALAYA	16.5	29.0
## 887	MEGHALAYA	16.0	27.0
## 888	PUDUCHERRY	26.5	26.5
## 889	PUDUCHERRY	27.0	27.0
## 890	PUDUCHERRY	26.5	26.5
## 891	PUDUCHERRY	29.0	29.0
## 892	PUNJAB	12.0	35.0
## 893	PUNJAB	12.0	34.0
## 894	PUNJAB	12.0	35.0
## 895	PUNJAB	10.0	32.0
## 896	PUNJAB	14.0	34.0
## 897	PUNJAB	11.0	30.0
## 898	PUNJAB	12.0	29.0
## 899	PUNJAB	12.0	36.0
## 900	PUNJAB	13.0	34.0
## 901	PUNJAB	14.0	34.0
## 902	PUNJAB	14.0	33.0
## 903	PUNJAB	14.0	34.0
## 904	PUNJAB	13.0	34.0
## 905	PUNJAB	13.0	36.0
## 906	PUNJAB	13.0	35.0
## 907	RAJASTHAN	18.5	24.0
## 908	RAJASTHAN	14.0	31.5
## 909	RAJASTHAN	18.0	29.0
## 910	TAMIL NADU	24.0	31.0
## 911	TAMIL NADU	22.0	34.0
## 912	TAMIL NADU	25.0	27.0
## 913	TAMIL NADU	25.0	29.0
## 914	TAMIL NADU	24.0	29.0
## 915	TAMIL NADU	25.0	26.0
## 916	TAMIL NADU	23.0	24.0
## 917	TAMIL NADU	25.0	29.0

## 918	TAMIL NADU	25.0	29.0
## 919	TAMIL NADU	24.0	25.0
## 920	TAMIL NADU	26.0	29.0
## 921	TAMIL NADU	25.0	29.0
## 922	TAMIL NADU	25.0	30.0
## 923	TAMIL NADU	25.0	25.0
## 924	TRIPURA	21.0	28.0
## 925	TRIPURA	21.0	27.0
## 926	TRIPURA	20.0	28.0
## 927	TRIPURA	20.5	28.0
## 928	TRIPURA	21.0	27.5
## 929	TRIPURA	21.0	28.0
## 930	TRIPURA	21.0	28.0
## 931	TRIPURA	20.5	27.5
## 932	TRIPURA	20.5	27.5
## 933	TRIPURA	21.5	28.5
## 934	TRIPURA	21.0	28.0
## 935	TRIPURA	20.5	27.5
## 936	TRIPURA	21.5	27.5
## 937	TRIPURA	21.0	28.0
## 938	TRIPURA	22.0	28.0
## 939	TRIPURA	21.5	27.5
## 940	TRIPURA	21.0	28.0
## 941	TRIPURA	21.0	28.0
## 942	TRIPURA	20.0	27.0
## 943	TRIPURA	21.0	27.5
## 944	TRIPURA	21.5	28.0
## 945	TRIPURA	21.5	27.5
## 946	TRIPURA	21.0	27.5
## 947	TRIPURA	21.0	27.0
## 948	TRIPURA	21.5	28.0
## 949	TRIPURA	21.0	28.0
## 950	State Name	Min	Max
## 951	ANDHRA PRADESH	25	30
## 952	ANDHRA PRADESH	24	32
## 953	ANDHRA PRADESH	24	33
## 954	ANDHRA PRADESH	24	31
## 955	ANDHRA PRADESH	22	33
## 956	ANDHRA PRADESH	24	29
## 957	ANDHRA PRADESH	25	27
## 958	ANDHRA PRADESH	24	28
## 959	ANDHRA PRADESH	25	25
## 960	ASSAM	18	32
## 961	ASSAM	18	30
## 962	ASSAM	20	32
## 963	ASSAM	19	32
## 964	ASSAM	19	29
## 965	ASSAM	17	29
## 966	ASSAM	17	29
## 967	ASSAM	18	32
## 968	ASSAM	18	30
## 969	ASSAM	20	26
## 970	ASSAM	24	25
## 971	ASSAM	20	26

## 972	ASSAM	21	25
## 973	ASSAM	19	25
## 974	ASSAM	20	28
## 975	ASSAM	20	30
## 976	ASSAM	24	33
## 977	ASSAM	17	31
## 978	ASSAM	20	25
## 979	ASSAM	19	30
## 980	ASSAM	19	32
## 981	ASSAM	19	34
## 982	BIHAR	16	28
## 983	BIHAR	17	32
## 984	BIHAR	16	28
## 985	BIHAR	19	29
## 986	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	28	32
## 987	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	29	33
## 988	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	29	30
## 989	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	28	32
## 990	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	28	33
## 991	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	28	32
## 992	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	28	32
## 993	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	29	32
## 994	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	29	32
## 995	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	26	27
## 996	GOA	26	32
## 997	GOA	28	32
## 998	GOA	27	31
## 999	GOA	26	30
## 1000	GOA	26	31
## 1001	GOA	27	30
## 1002	GOA	28	31
## 1003	GOA	27	31
## 1004	GOA	27	30
## 1005	GOA	27	30
## 1006	GOA	27	31
## 1007	GOA	27	30
## 1008	GOA	27	31
## 1009	GOA	28	30
## 1010	GOA	27	32
## 1011	GOA	26	31
## 1012	GOA	26	31
## 1013	GOA	28	31
## 1014	GOA	27	31
## 1015	GOA	28	30
## 1016	GOA	28	30
## 1017	GOA	26	30
## 1018	GOA	26	30
## 1019	GOA	26	30
## 1020	GOA	27	30
## 1021	GOA	28	32
## 1022	GUJARAT	24	35
## 1023	GUJARAT	22	30
## 1024	GUJARAT	22	28
## 1025	GUJARAT	22	30

## 1026	GUJARAT	28	32
## 1027	GUJARAT	22	30
## 1028	GUJARAT	18	30
## 1029	GUJARAT	26	32
## 1030	GUJARAT	26	31
## 1031	GUJARAT	24	34
## 1032	GUJARAT	24	35
## 1033	GUJARAT	25	31
## 1034	GUJARAT	28	32
## 1035	GUJARAT	24	32
## 1036	GUJARAT	24	30
## 1037	GUJARAT	22	33
## 1038	GUJARAT	22	33
## 1039	GUJARAT	26	27
## 1040	HARYANA	-	-
## 1041	HARYANA	-	-
## 1042	HARYANA	11	29
## 1043	HARYANA	14	29
## 1044	HARYANA	-	-
## 1045	HARYANA	-	-
## 1046	HARYANA	-	-
## 1047	HARYANA	-	-
## 1048	HARYANA	-	-
## 1049	HARYANA	-	-
## 1050	HIMACHAL\r\nPRADESH	20	33
## 1051	HIMACHAL\r\nPRADESH	20	29
## 1052	HIMACHAL\r\nPRADESH	25	25
## 1053	HIMACHAL\r\nPRADESH	21	26
## 1054	HIMACHAL\r\nPRADESH	20	28
## 1055	HIMACHAL\r\nPRADESH	8	18
## 1056	HIMACHAL\r\nPRADESH	9	20
## 1057	HIMACHAL\r\nPRADESH	9	23
## 1058	HIMACHAL\r\nPRADESH	9	24
## 1059	HIMACHAL\r\nPRADESH	13	25
## 1060	HIMACHAL\r\nPRADESH	13	26
## 1061	HIMACHAL\r\nPRADESH	16	19
## 1062	HIMACHAL\r\nPRADESH	3	13
## 1063	HIMACHAL\r\nPRADESH	6	16
## 1064	HIMACHAL\r\nPRADESH	11	23
## 1065	HIMACHAL\r\nPRADESH	16	23
## 1066	HIMACHAL\r\nPRADESH	16	22
## 1067	HIMACHAL\r\nPRADESH	9	16
## 1068	HIMACHAL\r\nPRADESH	8	15
## 1069	HIMACHAL\r\nPRADESH	10	16
## 1070	HIMACHAL\r\nPRADESH	17	19
## 1071	HIMACHAL\r\nPRADESH	-	-
## 1072	HIMACHAL\r\nPRADESH	-	-
## 1073	JHARKHAND	16	36
## 1074	JHARKHAND	18	30
## 1075	JHARKHAND	19	26
## 1076	JHARKHAND	17	29
## 1077	JHARKHAND	12	31
## 1078	JHARKHAND	21	27
## 1079	JHARKHAND	9	27

## 1080	JHARKHAND	9	26
## 1081	KARNATAKA	26	28
## 1082	KARNATAKA	25	33
## 1083	KARNATAKA	25	31
## 1084	KARNATAKA	24	29
## 1085	KARNATAKA	26	31
## 1086	KARNATAKA	25	31
## 1087	KARNATAKA	25	29
## 1088	KARNATAKA	24	28
## 1089	KARNATAKA	25	30
## 1090	KARNATAKA	25	29
## 1091	KARNATAKA	24	29
## 1092	KARNATAKA	22	28
## 1093	KARNATAKA	25	30
## 1094	KARNATAKA	26	31
## 1095	KERALA	26	32
## 1096	KERALA	24	31
## 1097	KERALA	22	33
## 1098	KERALA	24	30
## 1099	KERALA	21	23
## 1100	KERALA	26	32
## 1101	KERALA	27	31
## 1102	KERALA	26	29
## 1103	KERALA	26	30
## 1104	KERALA	27	31
## 1105	KERALA	26	32
## 1106	KERALA	25	31
## 1107	KERALA	27	32
## 1108	KERALA	20	23
## 1109	KERALA	25	31
## 1110	KERALA	26	29
## 1111	KERALA	20	22
## 1112	KERALA	26	29
## 1113	KERALA	20	24
## 1114	KERALA	27	31
## 1115	KERALA	22	24
## 1116	KERALA	22	31
## 1117	KERALA	22	31
## 1118	KERALA	26	30
## 1119	KERALA	28	30
## 1120	KERALA	27	31
## 1121	KERALA	27	30
## 1122	KERALA	27	31
## 1123	KERALA	25	29
## 1124	KERALA	26	29
## 1125	KERALA	26	29
## 1126	KERALA	20	23
## 1127	KERALA	20	23
## 1128	KERALA	20	31
## 1129	KERALA	22	32
## 1130	KERALA	20	31
## 1131	KERALA	22	23
## 1132	KERALA	22	24
## 1133	KERALA	23	31

## 1134	KERALA	25	31
## 1135	KERALA	24	29
## 1136	KERALA	24	29
## 1137	KERALA	24	28
## 1138	KERALA	25	30
## 1139	KERALA	24	29
## 1140	KERALA	24	30
## 1141	KERALA	27	30
## 1142	KERALA	27	29
## 1143	KERALA	26	29
## 1144	KERALA	27	29
## 1145	KERALA	27	29
## 1146	KERALA	26	29
## 1147	KERALA	25	29
## 1148	KERALA	25	30
## 1149	KERALA	26	32
## 1150	KERALA	26	32
## 1151	KERALA	26	31
## 1152	KERALA	28	32
## 1153	KERALA	26	31
## 1154	KERALA	26	30
## 1155	KERALA	27	31
## 1156	KERALA	27	30
## 1157	KERALA	22	33
## 1158	KERALA	20	34
## 1159	KERALA	24	28
## 1160	KERALA	23	31
## 1161	KERALA	20	23
## 1162	KERALA	20	24
## 1163	KERALA	27	30
## 1164	KERALA	25	28
## 1165	KERALA	25	30
## 1166	KERALA	26	32
## 1167	KERALA	27	32
## 1168	MADHYA PRADESH	20	30
## 1169	MADHYA PRADESH	20	32
## 1170	MADHYA PRADESH	20	26
## 1171	MADHYA PRADESH	27	28
## 1172	MADHYA PRADESH	18	25
## 1173	MADHYA PRADESH	17	27
## 1174	MADHYA PRADESH	19	25
## 1175	MADHYA PRADESH	17	26
## 1176	MADHYA PRADESH	20	25
## 1177	MADHYA PRADESH	25	28
## 1178	MADHYA PRADESH	26	27
## 1179	MAHARASHTRA	27	30
## 1180	MAHARASHTRA	27	30
## 1181	MAHARASHTRA	27	30
## 1182	MAHARASHTRA	28	35
## 1183	MAHARASHTRA	25	35
## 1184	MAHARASHTRA	26	30
## 1185	MAHARASHTRA	28	36
## 1186	MAHARASHTRA	27	30
## 1187	MAHARASHTRA	25	39

## 1188	MAHARASHTRA	24	30
## 1189	MAHARASHTRA	23	35
## 1190	MAHARASHTRA	25	34
## 1191	MAHARASHTRA	23	35
## 1192	MAHARASHTRA	20	32
## 1193	MAHARASHTRA	22	31
## 1194	MAHARASHTRA	31	32
## 1195	MAHARASHTRA	25	35
## 1196	MAHARASHTRA	23	35
## 1197	MAHARASHTRA	24	34
## 1198	MAHARASHTRA	28	35
## 1199	MAHARASHTRA	28	35
## 1200	MAHARASHTRA	27	35
## 1201	MAHARASHTRA	26	35
## 1202	MAHARASHTRA	28	35
## 1203	MAHARASHTRA	22	34
## 1204	MAHARASHTRA	25	34
## 1205	MAHARASHTRA	25	34
## 1206	MAHARASHTRA	25	34
## 1207	MAHARASHTRA	25	34
## 1208	MAHARASHTRA	25	27
## 1209	MAHARASHTRA	25	27
## 1210	MAHARASHTRA	25	27
## 1211	MAHARASHTRA	20	28
## 1212	MAHARASHTRA	18	28
## 1213	MAHARASHTRA	25	39
## 1214	MAHARASHTRA	25	38
## 1215	MANIPUR	17	30
## 1216	MANIPUR	17	30
## 1217	MANIPUR	16	29
## 1218	MANIPUR	19	29
## 1219	MANIPUR	3	30
## 1220	MANIPUR	19	31
## 1221	MANIPUR	17	30
## 1222	MANIPUR	18	31
## 1223	MANIPUR	20	29
## 1224	MANIPUR	20	29
## 1225	MANIPUR	19	29
## 1226	MANIPUR	11	28
## 1227	MANIPUR	22	28
## 1228	MANIPUR	19	28
## 1229	MANIPUR	19	28
## 1230	MANIPUR	23	28
## 1231	MANIPUR	23	27
## 1232	MANIPUR	21	29
## 1233	MANIPUR	20	28
## 1234	MANIPUR	21	28
## 1235	MANIPUR	18	29
## 1236	MANIPUR	19	32
## 1237	MANIPUR	20	28
## 1238	MANIPUR	18	27
## 1239	MANIPUR	19	29
## 1240	MANIPUR	27	29
## 1241	MANIPUR	20	28

## 1242	MANIPUR	19	28
## 1243	MANIPUR	21	27
## 1244	MANIPUR	19	28
## 1245	MANIPUR	20	28
## 1246	MEGHALAYA	16	27
## 1247	MEGHALAYA	10	26
## 1248	MEGHALAYA	8	22
## 1249	MEGHALAYA	19	31
## 1250	MEGHALAYA	19	31
## 1251	MEGHALAYA	16	26
## 1252	MEGHALAYA	16	22
## 1253	MEGHALAYA	16	23
## 1254	MEGHALAYA	14	24
## 1255	MEGHALAYA	9	22
## 1256	MEGHALAYA	10	22
## 1257	MEGHALAYA	18	28
## 1258	MEGHALAYA	14	22
## 1259	MEGHALAYA	13	27
## 1260	MEGHALAYA	13	24
## 1261	MEGHALAYA	13	26
## 1262	MEGHALAYA	14	24
## 1263	MEGHALAYA	15	25
## 1264	MEGHALAYA	12	22
## 1265	MEGHALAYA	13	24
## 1266	MEGHALAYA	13	23
## 1267	MEGHALAYA	15	26
## 1268	MEGHALAYA	15	28
## 1269	MEGHALAYA	16	28
## 1270	MEGHALAYA	13	27
## 1271	MEGHALAYA	8	31
## 1272	MEGHALAYA	17	31
## 1273	MEGHALAYA	7	31
## 1274	MEGHALAYA	19	30
## 1275	MEGHALAYA	18	31
## 1276	PUDUCHERRY	26	26
## 1277	PUDUCHERRY	26	26
## 1278	PUDUCHERRY	26	26
## 1279	PUDUCHERRY	26	26
## 1280	PUNJAB	10	30
## 1281	PUNJAB	12	29
## 1282	PUNJAB	12	29
## 1283	PUNJAB	11	29
## 1284	PUNJAB	11	28
## 1285	PUNJAB	10	29
## 1286	PUNJAB	9	29
## 1287	PUNJAB	12	30
## 1288	PUNJAB	12	29
## 1289	PUNJAB	12	29
## 1290	PUNJAB	11	37
## 1291	PUNJAB	10	36
## 1292	PUNJAB	13	29
## 1293	PUNJAB	13	29
## 1294	PUNJAB	11	29
## 1295	PUNJAB	-	-

## 1296	PUNJAB	-	-
## 1297	PUNJAB	-	-
## 1298	RAJASTHAN	17	30
## 1299	RAJASTHAN	18	22
## 1300	RAJASTHAN	15	30
## 1301	RAJASTHAN	19	29
## 1302	TAMIL NADU	22	28
## 1303	TAMIL NADU	25	27
## 1304	TAMIL NADU	25	29
## 1305	TAMIL NADU	26	32
## 1306	TAMIL NADU	25	28
## 1307	TAMIL NADU	3	27
## 1308	TAMIL NADU	24	30
## 1309	TAMIL NADU	24	24
## 1310	TAMIL NADU	25	30
## 1311	TAMIL NADU	24	28
## 1312	TAMIL NADU	26	31
## 1313	TAMIL NADU	25	27
## 1314	TAMIL NADU	25	28
## 1315	TAMIL NADU	25	27
## 1316	TRIPURA	22	29
## 1317	TRIPURA	23	30
## 1318	TRIPURA	23	31
## 1319	TRIPURA	22	30
## 1320	TRIPURA	22	29
## 1321	TRIPURA	22	28
## 1322	TRIPURA	22	29
## 1323	TRIPURA	22	30
## 1324	TRIPURA	22	29
## 1325	TRIPURA	22	29
## 1326	TRIPURA	22	28
## 1327	TRIPURA	23	30
## 1328	TRIPURA	23	29
## 1329	TRIPURA	22	29
## 1330	TRIPURA	22	28
## 1331	TRIPURA	22	29
## 1332	TRIPURA	22	30
## 1333	TRIPURA	22	29
## 1334	TRIPURA	22	29
## 1335	TRIPURA	22	29
## 1336	TRIPURA	22	28
## 1337	TRIPURA	22	28
## 1338	TRIPURA	22	28
## 1339	TRIPURA	22	28
## 1340	TRIPURA	22	28
## 1341	TRIPURA	22	28
## 1342	State Name	Min	Max
## 1343	ANDHRA PRADESH	23.0	32.0
## 1344	ANDHRA PRADESH	24.0	32.0
## 1345	ANDHRA PRADESH	24.0	32.0
## 1346	ANDHRA PRADESH	24.0	32.0
## 1347	ANDHRA PRADESH	22.0	28.0
## 1348	ANDHRA PRADESH	24.0	32.0
## 1349	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	26.2	30.0

## 1350	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	27.1	29.0
## 1351	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	27.1	30.0
## 1352	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	26.8	30.0
## 1353	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	26.4	30.5
## 1354	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	26.7	30.4
## 1355	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	27.1	30.0
## 1356	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	26.9	29.2
## 1357	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	20.0	27.0
## 1358	DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI	26.8	29.8
## 1359	GOA	28.0	32.0
## 1360	GOA	28.0	31.0
## 1361	GOA	27.0	29.0
## 1362	GOA	27.0	29.0
## 1363	GOA	26.0	31.0
## 1364	GOA	26.5	31.0
## 1365	GOA	28.0	32.0
## 1366	GOA	27.0	33.0
## 1367	GOA	27.0	30.0
## 1368	GOA	27.0	32.0
## 1369	GOA	27.0	31.0
## 1370	GOA	27.0	30.0
## 1371	GOA	27.0	30.0
## 1372	GOA	27.0	32.0
## 1373	GOA	28.0	30.0
## 1374	GOA	27.0	31.0
## 1375	GOA	27.0	31.0
## 1376	GOA	27.0	32.0
## 1377	GOA	28.0	31.0
## 1378	GOA	28.0	33.0
## 1379	GOA	27.0	31.0
## 1380	GOA	28.0	33.0
## 1381	GOA	27.0	30.0
## 1382	GOA	25.0	33.0
## 1383	GOA	28.0	30.0
## 1384	GOA	27.0	30.0
## 1385	GUJARAT	25.0	32.0
## 1386	GUJARAT	29.0	32.0
## 1387	GUJARAT	27.0	32.0
## 1388	GUJARAT	25.0	34.0
## 1389	GUJARAT	20.0	32.0
## 1390	GUJARAT	24.0	34.0
## 1391	GUJARAT	20.0	31.0
## 1392	GUJARAT	22.0	30.0
## 1393	GUJARAT	26.0	29.0
## 1394	GUJARAT	24.0	28.0
## 1395	GUJARAT	27.0	30.0
## 1396	GUJARAT	29.0	33.0
## 1397	GUJARAT	23.0	29.0
## 1398	GUJARAT	22.0	29.0
## 1399	GUJARAT	20.0	33.0
## 1400	GUJARAT	28.0	31.0
## 1401	GUJARAT	22.0	29.0
## 1402	HARYANA	23.0	33.0
## 1403	HARYANA	-	-

## 1404	HARYANA	-	-
## 1405	HARYANA	13.0	33.0
## 1406	HARYANA	22.0	32.0
## 1407	HARYANA	-	-
## 1408	HARYANA	-	-
## 1409	HARYANA	-	-
## 1410	HARYANA	25.0	33.0
## 1411	HIMACHAL\r\nPRADESH	15.0	32.0
## 1412	HIMACHAL\r\nPRADESH	19.0	25.0
## 1413	HIMACHAL\r\nPRADESH	17.0	26.0
## 1414	HIMACHAL\r\nPRADESH	18.0	26.0
## 1415	HIMACHAL\r\nPRADESH	16.0	26.0
## 1416	HIMACHAL\r\nPRADESH	19.0	25.0
## 1417	KARNATAKA	25.0	29.0
## 1418	KARNATAKA	25.0	28.0
## 1419	KARNATAKA	25.0	32.0
## 1420	KARNATAKA	25.0	32.0
## 1421	KARNATAKA	25.0	32.0
## 1422	KARNATAKA	25.0	32.0
## 1423	KERALA	20.0	25.0
## 1424	KERALA	21.0	23.0
## 1425	KERALA	21.0	25.0
## 1426	KERALA	22.0	30.0
## 1427	KERALA	25.0	30.0
## 1428	KERALA	25.0	29.0
## 1429	KERALA	26.0	27.0
## 1430	KERALA	25.0	33.0
## 1431	KERALA	21.0	30.0
## 1432	KERALA	28.0	33.0
## 1433	KERALA	28.0	30.0
## 1434	KERALA	28.0	31.0
## 1435	KERALA	26.0	34.0
## 1436	KERALA	28.0	31.0
## 1437	KERALA	26.0	31.0
## 1438	KERALA	22.0	24.0
## 1439	KERALA	21.0	24.0
## 1440	KERALA	22.0	24.0
## 1441	KERALA	28.0	31.0
## 1442	KERALA	27.0	32.0
## 1443	KERALA	20.0	23.0
## 1444	KERALA	29.0	32.0
## 1445	KERALA	25.0	29.0
## 1446	KERALA	26.0	28.0
## 1447	KERALA	26.0	27.0
## 1448	KERALA	27.0	29.0
## 1449	KERALA	23.0	28.0
## 1450	KERALA	26.0	29.0
## 1451	KERALA	23.0	30.0
## 1452	KERALA	24.0	31.0
## 1453	KERALA	25.0	29.0
## 1454	KERALA	25.0	30.0
## 1455	KERALA	26.0	33.0
## 1456	KERALA	26.0	32.0
## 1457	KERALA	26.0	28.0

## 1458	KERALA	28.0	30.0
## 1459	KERALA	28.0	30.0
## 1460	KERALA	27.0	33.0
## 1461	KERALA	27.0	28.0
## 1462	KERALA	24.0	34.0
## 1463	KERALA	28.0	29.0
## 1464	KERALA	25.0	29.0
## 1465	KERALA	27.0	29.0
## 1466	KERALA	27.0	29.0
## 1467	KERALA	26.0	27.0
## 1468	KERALA	26.0	28.0
## 1469	KERALA	21.0	23.0
## 1470	KERALA	26.0	30.0
## 1471	KERALA	27.0	32.0
## 1472	KERALA	27.0	30.0
## 1473	KERALA	27.0	32.0
## 1474	KERALA	26.0	29.0
## 1475	KERALA	27.0	31.0
## 1476	KERALA	26.0	35.0
## 1477	KERALA	29.0	31.0
## 1478	KERALA	26.0	33.0
## 1479	KERALA	26.0	30.0
## 1480	KERALA	27.0	31.0
## 1481	KERALA	28.0	30.0
## 1482	KERALA	27.0	32.0
## 1483	KERALA	25.0	29.0
## 1484	KERALA	23.0	28.0
## 1485	KERALA	23.0	31.0
## 1486	KERALA	25.0	30.0
## 1487	KERALA	27.0	28.0
## 1488	KERALA	25.0	28.0
## 1489	KERALA	22.0	24.0
## 1490	KERALA	22.0	24.0
## 1491	KERALA	27.0	29.0
## 1492	KERALA	26.0	29.0
## 1493	KERALA	26.0	27.0
## 1494	MAHARASHTRA	28.0	36.0
## 1495	MAHARASHTRA	19.0	34.0
## 1496	MAHARASHTRA	28.0	34.0
## 1497	MAHARASHTRA	18.0	33.0
## 1498	MAHARASHTRA	21.0	28.0
## 1499	MAHARASHTRA	-	-
## 1500	MAHARASHTRA	28.0	36.0
## 1501	MAHARASHTRA	27.0	36.0
## 1502	MAHARASHTRA	27.0	36.0
## 1503	MAHARASHTRA	27.0	36.0
## 1504	MAHARASHTRA	25.0	28.0
## 1505	MAHARASHTRA	22.0	34.0
## 1506	MAHARASHTRA	26.0	30.0
## 1507	MAHARASHTRA	26.0	30.0
## 1508	MAHARASHTRA	28.0	30.0
## 1509	MAHARASHTRA	28.0	30.0
## 1510	MAHARASHTRA	27.0	30.0
## 1511	MAHARASHTRA	27.0	30.0

## 1512	MAHARASHTRA	26.0	30.0
## 1513	MAHARASHTRA	21.0	38.0
## 1514	MAHARASHTRA	25.0	34.0
## 1515	MAHARASHTRA	26.0	35.0
## 1516	MAHARASHTRA	26.0	35.0
## 1517	MAHARASHTRA	26.0	40.0
## 1518	MAHARASHTRA	26.0	40.0
## 1519	MAHARASHTRA	19.0	24.0
## 1520	MAHARASHTRA	27.0	27.0
## 1521	MAHARASHTRA	27.0	27.0
## 1522	MAHARASHTRA	27.0	27.0
## 1523	MAHARASHTRA	20.0	25.0
## 1524	MAHARASHTRA	25.0	38.0
## 1525	MAHARASHTRA	25.0	38.0
## 1526	MAHARASHTRA	25.0	38.0
## 1527	MANIPUR	18.0	26.0
## 1528	MANIPUR	19.0	28.0
## 1529	MANIPUR	20.0	29.0
## 1530	MANIPUR	19.0	28.0
## 1531	MANIPUR	21.0	29.0
## 1532	MANIPUR	20.0	28.0
## 1533	MANIPUR	20.0	29.0
## 1534	MANIPUR	21.0	28.0
## 1535	MANIPUR	19.0	27.0
## 1536	MANIPUR	16.0	28.0
## 1537	MANIPUR	20.0	28.0
## 1538	MANIPUR	21.0	28.0
## 1539	MANIPUR	18.0	28.0
## 1540	MANIPUR	21.0	28.0
## 1541	MANIPUR	18.4	28.0
## 1542	MANIPUR	20.0	27.0
## 1543	MANIPUR	21.0	29.0
## 1544	MANIPUR	17.0	28.0
## 1545	MANIPUR	18.0	28.0
## 1546	MANIPUR	20.0	28.0
## 1547	MANIPUR	20.0	28.0
## 1548	MANIPUR	17.0	28.0
## 1549	MANIPUR	21.0	29.0
## 1550	MANIPUR	23.0	30.0
## 1551	MANIPUR	24.0	28.0
## 1552	MANIPUR	20.0	27.0
## 1553	MANIPUR	19.0	29.0
## 1554	MANIPUR	19.0	28.0
## 1555	MANIPUR	19.0	28.0
## 1556	MANIPUR	19.0	27.0
## 1557	MANIPUR	19.0	28.0
## 1558	MANIPUR	0.6	28.0
## 1559	MANIPUR	21.0	27.0
## 1560	MANIPUR	20.0	28.0
## 1561	MANIPUR	19.0	28.0
## 1562	MANIPUR	19.0	28.0
## 1563	MEGHALAYA	18.1	30.0
## 1564	MEGHALAYA	18.0	30.0
## 1565	MEGHALAYA	18.0	29.0

## 1566	MEGHALAYA	20.0	29.0
## 1567	MEGHALAYA	14.6	23.0
## 1568	MEGHALAYA	13.2	31.0
## 1569	MEGHALAYA	13.0	24.0
## 1570	MEGHALAYA	14.0	22.0
## 1571	MEGHALAYA	11.0	21.0
## 1572	MEGHALAYA	12.0	21.0
## 1573	MEGHALAYA	17.5	28.0
## 1574	MEGHALAYA	11.0	21.0
## 1575	MEGHALAYA	15.0	22.0
## 1576	MEGHALAYA	13.0	20.0
## 1577	MEGHALAYA	14.0	21.0
## 1578	MEGHALAYA	14.0	22.0
## 1579	MEGHALAYA	15.0	24.0
## 1580	MEGHALAYA	14.0	24.0
## 1581	MEGHALAYA	14.0	24.0
## 1582	MEGHALAYA	13.0	22.0
## 1583	MEGHALAYA	13.0	22.0
## 1584	MEGHALAYA	14.0	22.0
## 1585	MEGHALAYA	15.0	26.1
## 1586	MEGHALAYA	15.0	24.1
## 1587	MIZORAM	18.0	27.0
## 1588	MIZORAM	16.0	28.0
## 1589	MIZORAM	13.0	27.0
## 1590	MIZORAM	16.0	26.0
## 1591	MIZORAM	15.0	27.0
## 1592	MIZORAM	19.0	30.0
## 1593	MIZORAM	18.0	27.0
## 1594	MIZORAM	18.0	25.0
## 1595	MIZORAM	18.0	29.0
## 1596	MIZORAM	18.0	30.0
## 1597	MIZORAM	19.0	29.0
## 1598	MIZORAM	10.0	19.0
## 1599	MIZORAM	18.0	29.0
## 1600	ODISHA	20.0	30.0
## 1601	ODISHA	22.0	29.0
## 1602	ODISHA	22.0	31.0
## 1603	ODISHA	13.0	30.0
## 1604	ODISHA	14.0	28.0
## 1605	ODISHA	15.0	29.0
## 1606	ODISHA	12.0	26.0
## 1607	ODISHA	15.0	31.0
## 1608	ODISHA	13.0	31.0
## 1609	ODISHA	15.0	32.0
## 1610	ODISHA	14.0	32.0
## 1611	ODISHA	15.0	32.0
## 1612	ODISHA	12.0	33.0
## 1613	ODISHA	15.0	32.0
## 1614	PUDUCHERRY	24.5	28.2
## 1615	PUDUCHERRY	27.0	27.0
## 1616	PUDUCHERRY	26.5	26.5
## 1617	PUDUCHERRY	27.0	27.0
## 1618	PUDUCHERRY	29.0	29.0
## 1619	PUNJAB	10.0	36.0

## 1620		PUNJAB	10.0	37.0		
## 1621		PUNJAB	14.0	32.0		
## 1622		PUNJAB	13.0	30.0		
## 1623		PUNJAB	13.0	32.0		
## 1624		PUNJAB	14.0	38.0		
## 1625		PUNJAB	12.0	37.0		
## 1626		PUNJAB	18.2	33.0		
## 1627		PUNJAB	13.0	36.0		
## 1628		PUNJAB	10.0	36.0		
## 1629		PUNJAB	22.0	33.0		
## 1630		PUNJAB	25.0	32.0		
## 1631		PUNJAB	13.0	31.0		
## 1632		PUNJAB	15.0	30.0		
## 1633		PUNJAB	11.0	38.0		
## 1634		PUNJAB	13.0	32.0		
## 1635		PUNJAB	11.0	36.0		
## 1636		TAMIL NADU	23.0	26.0		
## 1637		TAMIL NADU	24.0	28.0		
## 1638		TAMIL NADU	23.0	28.0		
## 1639		TAMIL NADU	25.0	28.0		
## 1640		TAMIL NADU	25.0	29.0		
## 1641		TAMIL NADU	24.0	27.0		
## 1642		TAMIL NADU	26.0	29.0		
## 1643		TAMIL NADU	25.0	27.0		
## 1644		TAMIL NADU	23.0	29.0		
## 1645		TAMIL NADU	25.0	28.0		
## 1646		TAMIL NADU	25.0	29.0		
## 1647		TAMIL NADU	25.0	29.0		
## 1648		TAMIL NADU	25.0	35.0		
## 1649		TAMIL NADU	25.0	28.0		
## 1650		TRIPURA	18.0	29.0		
## 1651		TRIPURA	18.5	29.0		
## 1652		TRIPURA	18.0	29.0		
## 1653		TRIPURA	19.0	31.0		
## 1654		TRIPURA	18.5	29.0		
## 1655		TRIPURA	19.0	29.0		
## 1656		TRIPURA	17.5	28.5		
## 1657		TRIPURA	18.0	27.5		
## 1658		TRIPURA	19.0	30.0		
## 1659		TRIPURA	19.5	31.0		
## 1660		TRIPURA	19.0	29.5		
## 1661		TRIPURA	19.5	31.0		
## 1662		TRIPURA	18.0	29.0		
## 1663		TRIPURA	18.0	29.0		
## 1664		TRIPURA	18.5	28.5		
## 1665		TRIPURA	18.5	28.0		
##	DissolvedO2Min	DissolvedO2Max	pHMin	pHMax	ConductivityMin	ConductivityMax
## 1	6.0	8.0	6.6	7.8	369	640
## 2	6.2	8.2	6.2	8.2	339	502
## 3	6.0	16.9	7.8	8.4	143	233
## 4	4.2	6.6	7.3	8.6	163	58350
## 5	4.3	6.2	7.3	8.1	1867	66950
## 6	5.2	6.8	7.2	8.4	1823	65980
## 7	4.6	12.8	7.4	8.9	163	418

## 8	5.2	6.9	7.9	8.4	199	318
## 9	4.5	6.6	7.9	8.3	193	353
## 10	5.6	7.3	7.9	8.7	156	256
## 11	4.4	8.6	7.4	8.2	301	45740
## 12	4.9	7.6	6.8	8.2	6755	57760
## 13	4.5	7.8	6.9	8.0	1485	59470
## 14	4.2	7.5	6.0	7.8	7	26350
## 15	4.3	7.8	5.9	7.5	116	47730
## 16	6.4	7.7	6.1	7.3	40	93
## 17	6.3	8.3	6.0	7.2	46	168
## 18	6.0	8.4	6.0	7.7	61	207
## 19	6.8	7.6	6.1	7.6	42	111
## 20	4.9	8.1	5.8	9.2	57	174
## 21	5.7	7.4	5.8	7.1	49	138
## 22	6.7	7.7	6.1	7.4	41	108
## 23	4.9	7.0	5.9	7.3	80	231
## 24	7.0	7.8	6.0	7.3	50	478
## 25	2.8	9.6	7.1	8.3	37020	65120
## 26	2.7	8.1	6.2	7.8	159	16380
## 27	6.0	7.8	6.3	7.6	5	4770
## 28	5.6	12.0	5.6	7.5	76	2432
## 29	4.6	7.6	6.2	7.8	275	58190
## 30	4.6	7.8	7.3	8.2	4669	75840
## 31	0.7	7.6	6.0	8.0	125	778
## 32	3.9	9.9	6.1	8.8	38	43800
## 33	3.9	7.0	7.6	8.4	9666	79480
## 34	3.9	7.7	7.5	8.0	7915	83240
## 35	5.5	6.8	7.4	8.3	326	67670
## 36	4.8	6.4	7.1	7.9	298	10880
## 37	6.8	7.5	7.7	8.0	222	344
## 38	5.6	6.5	7.3	8.0	352	50320
## 39	0.9	5.2	7.1	8.2	778	2280
## 40	6.5	7.2	7.4	8.0	78	53810
## 41	2.9	3.1	7.3	8.2	1090	6110
## 42	3.6	6.0	5.9	8.0	704	3495
## 43	4.7	6.4	7.1	8.2	433	1738
## 44	5.2	6.8	7.2	8.1	358	7633
## 45	5.2	6.5	7.4	8.2	398	64670
## 46	1.5	4.4	7.1	8.3	538	3360
## 47	3.2	5.3	7.2	8.4	740	2600
## 48	2.4	6.0	7.2	7.7	560	2280
## 49	5.0	6.7	7.2	8.4	992	28800
## 50	0.4	6.9	6.5	8.1	495	2140
## 51	4.1	7.9	7.5	8.3	300	646
## 52	2.6	4.2	6.7	8.5	564	2640
## 53	2.7	4.2	6.7	8.7	561	2910
## 54	2.3	6.8	4.9	7.9	327	3160
## 55	6.9	8.2	7.0	8.2	283	574
## 56	2.4	4.1	6.4	8.2	540	3140
## 57	7.2	7.9	7.5	8.3	528	764
## 58	6.9	8.0	7.4	8.1	496	716
## 59	3.2	7.7	7.2	8.3	466	768
## 60	6.6	8.3	7.2	8.2	486	2361
## 61	2.1	5.7	2.8	8.0	92	1980

## 62	6.4	7.3	6.9	8.3	482	987
## 63	6.8	7.8	6.8	7.8	60	100
## 64	6.2	6.7	6.7	7.8	126	366
## 65	6.2	6.7	6.7	7.8	116	163
## 66	7.2	7.9	6.8	8.4	52	92
## 67	6.2	7.0	6.7	7.6	124	152
## 68	6.0	7.0	6.7	7.6	131	168
## 69	4.1	7.0	6.4	7.6	100	39410
## 70	4.7	7.8	6.4	7.6	30	63
## 71	4.2	8.0	6.6	8.3	35	90
## 72	4.6	8.7	6.5	7.8	55	41000
## 73	6.5	7.8	6.8	7.6	92	131
## 74	5.5	6.9	6.5	7.9	54	700
## 75	3.9	7.6	6.1	7.0	47	251
## 76	1.5	1.9	6.6	7.3	600	1100
## 77	1.5	7.1	5.6	6.7	42	63
## 78	3.9	7.1	6.3	7.1	33	64
## 79	6.1	7.3	6.4	7.7	40	92
## 80	3.5	8.0	5.4	8.3	58	190
## 81	1.0	6.1	5.3	7.2	45	83
## 82	6.1	7.2	6.7	7.5	110	130
## 83	5.8	7.2	6.5	8.2	58	239
## 84	4.8	6.9	6.0	6.8	100	470
## 85	5.5	7.3	6.6	7.5	110	144
## 86	6.0	6.6	6.2	6.8	76	120
## 87	4.5	7.1	6.8	7.7	110	158
## 88	1.7	6.7	5.9	7.1	49	73
## 89	5.8	7.0	6.5	7.2	62	280
## 90	5.1	8.0	6.5	7.3	20	34
## 91	5.3	7.8	6.1	7.3	28	98
## 92	4.6	8.1	5.7	7.5	94	47000
## 93	3.2	7.0	6.2	7.9	95	50000
## 94	6.7	8.1	5.4	8.1	47	98
## 95	7.0	8.4	6.4	7.7	51	80
## 96	0.6	5.5	6.2	7.2	118	9700
## 97	6.8	7.3	6.6	7.0	70	162
## 98	6.4	7.2	6.3	6.9	77	120
## 99	6.4	7.0	6.0	6.8	89	130
## 100	5.1	7.2	6.6	7.5	12230	16820
## 101	5.4	7.1	6.9	7.5	124	160
## 102	5.1	8.0	6.9	7.6	35	146
## 103	2.4	7.4	6.6	7.8	85	230
## 104	4.9	8.5	6.8	7.8	91	240
## 105	3.1	6.4	6.3	7.4	48	30750
## 106	5.1	8.0	6.5	7.5	58	143
## 107	1.1	4.3	6.8	8.0	767	50980
## 108	2.4	7.8	6.8	7.8	329	40360
## 109	3.6	8.3	6.2	7.2	57	1200
## 110	3.5	9.0	6.1	7.3	42	211
## 111	6.5	8.7	6.3	7.5	36	99
## 112	2.5	8.4	5.9	7.7	118	59900
## 113	3.9	8.1	5.8	7.4	32	120
## 114	3.6	7.7	5.1	7.7	38	3549
## 115	3.9	8.3	6.7	8.3	75	40000

## 116	6.9	8.2	6.1	7.8	43	82
## 117	4.5	7.5	6.1	7.8	252	51000
## 118	5.9	8.5	6.1	7.7	53	84
## 119	4.3	7.7	6.4	7.9	105	52110
## 120	3.7	8.2	6.4	8.2	70	150
## 121	4.8	8.0	6.2	7.4	50	32860
## 122	0.5	10.3	7.1	8.2	355	955
## 123	6.6	9.4	6.6	8.0	10	848
## 124	6.8	8.8	6.2	8.3	175	340
## 125	4.7	6.6	6.3	7.2	45	148
## 126	2.6	9.2	6.4	8.1	23	32500
## 127	1.4	6.1	6.3	7.0	56	170
## 128	5.1	7.9	6.5	7.5	48	19000
## 129	0.0	2.2	6.0	7.1	80	189
## 130	0.0	3.6	6.1	7.3	71	132
## 131	5.6	7.7	6.5	7.5	97	141
## 132	5.6	7.0	6.4	7.6	86	178
## 133	4.3	7.2	7.2	8.2	187	365
## 134	5.6	7.8	6.8	8.0	28	114
## 135	6.5	7.4	7.1	7.8	102	164
## 136	6.4	7.7	6.3	7.3	89	128
## 137	4.3	7.2	6.4	7.3	34	70
## 138	7.1	8.9	6.4	7.3	42	99
## 139	1.6	8.0	6.2	7.9	86	50000
## 140	BDL	7.3	6.4	7.8	121	15450
## 141	6.0	7.4	7.0	8.0	139	236
## 142	6.2	7.5	6.3	8.1	94	168
## 143	6.0	7.3	6.7	7.7	62	193
## 144	6.3	7.4	6.8	8.1	71	287
## 145	6.2	7.5	6.7	8.3	76	655
## 146	6.1	7.2	6.6	8.3	68	196
## 147	6.3	7.4	7.2	8.0	110	170
## 148	5.7	7.1	6.5	8.3	62	551
## 149	BDL	5.6	6.6	7.6	297	2139
## 150	5.0	7.2	6.7	8.3	109	21540
## 151	5.4	7.1	6.8	8.0	162	41360
## 152	6.0	7.4	7.0	8.1	70	147
## 153	5.4	7.3	6.5	8.1	70	112
## 154	6.4	7.7	6.5	8.3	67	103
## 155	4.4	6.6	7.8	8.4	367	613
## 156	3.5	6.0	6.7	7.9	4124	69610
## 157	6.4	7.6	6.9	7.9	63	84
## 158	6.5	7.0	7.2	8.8	83	1008
## 159	3.3	7.0	6.1	7.7	113	59681
## 160	5.6	7.3	6.6	7.4	71	195
## 161	6.0	7.4	6.7	7.7	72	194
## 162	5.8	7.5	6.6	8.0	69	299
## 163	5.0	7.2	6.5	8.0	62	187
## 164	6.5	7.5	6.8	8.2	117	2328
## 165	6.5	7.1	6.5	7.9	104	425
## 166	5.8	7.0	6.7	7.9	109	24440
## 167	5.0	7.0	6.5	8.0	98	32500
## 168	5.7	7.0	6.7	7.8	104	36750
## 169	6.3	7.6	7.1	8.3	135	1170

## 170	6.2	7.6	7.5	8.3	134	1180
## 171	6.4	7.5	7.4	8.3	133	1202
## 172	6.4	7.4	7.3	8.0	113	276
## 173	6.8	7.3	7.0	7.9	107	270
## 174	5.8	7.1	6.8	8.2	68	234
## 175	6.7	7.1	6.7	8.0	75	181
## 176	5.4	6.3	7.2	7.4	320	368
## 177	5.9	6.1	7.1	7.2	390	532
## 178	5.8	6.1	7.3	7.8	565	576
## 179	5.9	6.7	5.9	6.9	290	373
## 180	2.1	6.2	6.5	6.8	590	694
## 181	2.3	4.4	6.6	7.1	610	630
## 182	3.9	6.4	6.7	6.9	420	568
## 183	4.3	6.8	7.0	7.3	250	396
## 184	6.2	7.1	7.2	7.4	300	495
## 185	7.2	7.6	7.2	7.4	196	350
## 186	6.7	7.9	7.0	7.6	135	301
## 187	6.0	6.5	6.1	7.8	155	251
## 188	7.4	7.9	7.3	7.5	3	275
## 189	7.1	7.8	7.1	7.6	169	280
## 190	7.3	7.9	6.2	7.4	150	290
## 191	6.4	6.9	6.2	6.7	450	705
## 192	6.2	6.9	6.9	9.0	120	212
## 193	6.1	6.5	6.8	7.7	430	689
## 194	6.0	6.3	7.1	7.8	230	582
## 195	6.2	6.3	7.2	7.9	250	526
## 196	6.3	6.5	6.2	7.8	270	510
## 197	6.6	7.2	6.5	7.4	200	306
## 198	7.3	8.2	7.1	7.6	200	380
## 199	7.2	8.4	7.1	7.4	180	230
## 200	7.1	7.9	6.3	7.4	100	175
## 201	7.2	7.8	7.1	7.9	100	180
## 202	6.4	6.9	7.1	7.4	125	190
## 203	7.2	7.9	2.4	7.6	95	169
## 204	7.2	7.9	6.3	7.2	140	340
## 205	6.3	6.5	7.3	7.6	125	259
## 206	7.1	7.3	7.1	7.2	150	303
## 207	6.1	7.2	6.3	6.9	350	583
## 208	5.9	7.2	6.1	7.2	355	552
## 209	5.6	7.6	7.0	7.4	55	102
## 210	4.2	7.9	2.8	3.7	330	980
## 211	5.6	7.0	6.7	7.3	50	97
## 212	6.9	8.1	7.1	8.0	58	143
## 213	6.8	8.3	7.1	7.9	58	140
## 214	6.2	7.5	7.0	7.2	48	65
## 215	5.9	8.5	6.8	7.6	36	69
## 216	6.3	8.4	7.1	7.7	40	69
## 217	5.6	7.6	6.7	7.2	43	96
## 218	5.8	8.0	6.7	7.5	36	71
## 219	6.9	8.2	7.0	7.5	32	57
## 220	7.0	7.8	7.4	8.0	45	132
## 221	6.4	8.5	6.5	7.3	33	63
## 222	0.3	8.0	6.8	7.5	33	83
## 223	1.0	6.8	6.8	7.8	136	620

## 224	3.0	6.6	7.2	7.6	180	335
## 225	1.2	7.8	7.2	7.3	220	520
## 226	0.4	5.0	6.8	7.5	300	790
## 227	6.4	7.6	6.8	7.2	30	444
## 228	0.5	5.4	6.6	7.2	149	589
## 229	2.5	7.0	7.0	7.3	110	354
## 230	6.0	8.6	7.1	7.5	36	55
## 231	3.9	7.6	2.8	4.4	370	1420
## 232	6.3	8.1	6.7	7.6	68	395
## 233	5.8	8.4	5.0	6.5	32	69
## 234	6.8	7.6	7.3	7.9	68	155
## 235	6.9	7.5	7.4	8.0	65	162
## 236	9.4	9.4	8.5	8.5	6190	6190
## 237	8.5	8.5	8.3	8.3	10160	10160
## 238	7.1	7.1	8.4	8.4	6960	6960
## 239	6.6	6.6	7.1	7.1	324	324
## 240	3.8	6.4	7.4	8.2	320	664
## 241	BDL	4.3	7.4	8.2	337	1833
## 242	2.2	5.0	7.0	7.9	442	990
## 243	1.0	4.4	7.3	8.0	466	1533
## 244	3.4	5.2	7.3	7.9	8	783
## 245	2.1	4.2	7.3	8.0	422	977
## 246	1.8	4.0	7.4	8.2	448	1304
## 247	1.2	4.0	7.3	8.0	416	1510
## 248	2.6	4.1	7.2	8.0	350	944
## 249	1.1	4.2	7.3	8.2	410	1368
## 250	0.6	4.2	7.7	8.3	351	1490
## 251	0.8	4.9	7.6	8.1	316	1485
## 252	0.0	0.0	7.2	7.5	1551	2346
## 253	4.7	5.7	7.6	8.4	270	680
## 254	4.5	5.9	7.7	8.3	280	930
## 255	4.9	5.6	7.8	8.4	300	940
## 256	5.5	7.5	6.6	7.6	64	421
## 257	6.2	7.8	7.1	8.1	98	231
## 258	6.0	7.2	7.2	8.2	33	1161
## 259	6.5	7.5	6.7	8.3	115	374
## 260	5.2	7.7	7.0	7.8	88	167
## 261	6.0	7.2	6.8	7.7	76	128
## 262	6.0	7.5	7.7	8.5	549	57900
## 263	6.2	7.6	7.4	8.8	188	546
## 264	6.3	9.8	7.0	7.7	85	139
## 265	6.2	8.2	7.5	8.4	128	357
## 266	5.9	7.2	7.2	8.2	121	252
## 267	6.0	7.5	7.3	8.3	145	264
## 268	5.3	7.0	7.3	8.5	91	197
## 269	5.7	7.0	6.8	8.0	92	194
## 270	6.6	8.4	6.9	8.0	92	195
## 271	4.8	6.3	6.7	7.9	81	168
## 272	3.8	5.4	6.2	7.5	92	184
## 273	3.9	6.1	6.4	8.0	101	189
## 274	4.6	5.9	6.6	7.6	105	178
## 275	5.3	6.6	7.4	8.2	101	215
## 276	6.9	8.0	7.4	8.9	108	191
## 277	7.3	8.7	7.0	8.0	96	198

## 278	4.1	5.3	6.4	7.6	103	195
## 279	7.3	8.5	7.3	8.7	108	199
## 280	7.1	8.0	6.0	7.0	72	143
## 281	6.5	7.8	7.1	8.2	92	179
## 282	5.1	6.8	7.4	8.5	65	157
## 283	7.9	9.2	8.1	8.9	72	185
## 284	6.1	7.2	6.3	7.5	95	181
## 285	5.3	7.0	6.5	7.9	81	187
## 286	5.9	7.4	7.6	8.7	89	183
## 287	6.5	8.0	5.6	7.3	82	157
## 288	3.6	5.4	6.4	7.4	84	174
## 289	5.3	6.8	6.4	7.3	82	177
## 290	5.9	7.6	6.8	7.9	83	179
## 291	6.1	7.3	6.8	7.9	88	187
## 292	8.5	10.1	6.6	7.8	62	167
## 293	3.8	5.0	6.3	7.7	73	189
## 294	5.2	7.2	6.0	8.0	277	1342
## 295	5.1	7.2	6.9	8.0	248	7526
## 296	5.1	6.7	7.2	8.2	163	290
## 297	3.8	6.6	7.1	8.1	547	10794
## 298	4.1	6.8	7.3	8.4	7650	39705
## 299	4.6	7.2	6.9	7.9	6180	59212
## 300	4.9	6.5	7.1	8.4	207	374
## 301	4.8	6.8	7.4	8.4	184	385
## 302	5.2	7.0	7.4	8.1	199	411
## 303	5.1	7.0	7.7	8.5	182	304
## 304	4.5	6.6	7.0	8.5	288	5042
## 305	4.7	8.0	6.6	7.6	386	53780
## 306	4.8	8.9	6.9	7.9	49	52720
## 307	4.9	7.6	6.4	8.1	87	23360
## 308	4.7	7.7	6.5	7.4	40	40130
## 309	5.0	7.8	6.8	7.9	55	152
## 310	6.5	8.4	6.5	8.0	54	198
## 311	5.8	8.1	6.4	7.7	55	227
## 312	6.5	7.9	6.3	8.1	48	221
## 313	6.7	8.1	6.2	7.9	59	213
## 314	5.5	7.8	6.3	8.3	45	190
## 315	5.2	8.1	6.5	8.4	44	228
## 316	4.6	7.7	6.0	7.8	72	1196
## 317	5.9	7.8	6.5	7.6	47	220
## 318	4.8	8.0	7.0	8.1	56	55620
## 319	2.3	5.1	6.5	7.4	184	11890
## 320	4.8	7.9	6.6	8.2	66	7108
## 321	5.2	13.0	6.1	8.8	48	411
## 322	4.4	7.9	6.6	7.9	116	46920
## 323	4.0	7.8	6.5	7.7	115	53770
## 324	0.3	3.6	6.6	7.4	133	463
## 325	3.7	6.7	6.7	7.6	30	35490
## 326	4.7	6.5	6.7	8.0	16210	55000
## 327	5.1	6.5	6.7	7.9	20170	54680
## 328	5.2	8.5	6.4	7.8	3007	51100
## 329	5.7	7.1	7.1	8.6	402	34210
## 330	4.0	6.6	7.0	8.1	340	15050
## 331	6.9	7.1	7.1	8.1	282	395

## 332	3.8	6.4	7.1	7.9	264	39570
## 333	0.0	3.0	7.3	7.7	734	2350
## 334	5.3	6.8	7.1	8.0	380	24370
## 335	0.0	0.6	7.0	8.6	1524	6190
## 336	4.6	6.9	7.1	7.9	560	4060
## 337	5.0	6.9	7.0	8.1	335	1679
## 338	5.9	7.2	6.9	8.2	410	12470
## 339	6.4	7.1	7.2	8.4	465	35550
## 340	0.0	3.4	6.9	8.1	799	1684
## 341	0.0	2.4	7.3	8.0	1390	3400
## 342	0.0	5.8	7.2	7.9	530	2332
## 343	3.2	6.5	7.0	8.1	556	6410
## 344	0.0	4.1	6.5	8.0	1283	1970
## 345	6.9	8.3	7.8	8.5	379	1260
## 346	6.4	8.4	7.9	8.9	231	235
## 347	2.4	5.0	7.7	8.2	1190	6764
## 348	3.4	4.2	7.7	8.6	895	1396
## 349	3.2	4.9	7.5	8.6	685	1490
## 350	5.4	5.4	6.8	8.0	541	2250
## 351	4.2	4.2	7.6	8.2	312	581
## 352	3.0	4.1	7.5	8.3	980	6670
## 353	6.9	8.2	6.5	8.0	53	117
## 354	6.0	8.0	6.5	8.1	47	97
## 355	6.2	7.8	6.9	7.2	112	162
## 356	6.0	7.9	6.8	7.2	120	158
## 357	1.2	6.9	5.9	7.2	56	26600
## 358	5.3	8.0	6.4	6.9	32	55
## 359	5.2	8.6	6.7	8.4	56	88
## 360	5.1	7.8	6.6	7.6	60	50800
## 361	6.2	7.4	6.9	7.4	80	108
## 362	5.4	7.1	7.0	7.6	53	74
## 363	5.6	8.2	6.3	6.9	41	410
## 364	0.0	6.8	6.2	6.7	662	909
## 365	4.5	6.4	6.1	7.4	40	74
## 366	4.9	7.4	6.2	6.8	40	77
## 367	6.2	7.4	6.7	7.3	45	66
## 368	5.0	7.9	6.8	7.8	55	360
## 369	2.6	6.2	6.1	6.9	65	95
## 370	5.3	6.8	7.1	7.4	90	136
## 371	3.7	7.9	6.7	7.6	79	240
## 372	3.9	6.8	6.1	6.9	196	255
## 373	5.6	6.8	6.5	6.8	108	130
## 374	5.6	6.4	6.3	6.7	88	130
## 375	6.0	6.9	7.0	7.8	100	120
## 376	2.2	7.3	5.1	7.1	56	82
## 377	4.3	8.0	6.4	7.2	68	132
## 378	7.7	9.3	6.5	7.9	23	45
## 379	5.4	8.2	6.1	7.2	25	81
## 380	4.2	8.2	6.5	7.9	76	39800
## 381	4.3	7.9	6.6	7.4	110	48000
## 382	7.3	8.0	6.7	8.3	40	90
## 383	7.1	8.4	6.6	7.8	52	80
## 384	1.5	6.1	5.3	7.2	91	14830
## 385	5.6	7.7	6.4	6.9	95	165

## 386	6.6	7.5	6.4	6.8	100	140
## 387	5.2	6.9	6.4	6.9	122	160
## 388	5.8	6.9	7.0	7.4	9790	13106
## 389	6.0	6.8	6.7	7.4	110	132
## 390	4.0	6.9	6.3	7.1	46	144
## 391	2.0	9.6	6.2	7.3	108	183
## 392	1.5	8.9	6.1	7.4	108	168
## 393	3.1	7.1	6.4	7.5	75	23350
## 394	5.0	8.6	6.5	7.3	60	141
## 395	1.7	5.3	6.5	8.3	336	54800
## 396	4.1	7.6	6.6	7.7	199	44800
## 397	3.3	8.1	6.4	7.6	59	108
## 398	5.2	8.3	6.3	7.7	50	110
## 399	6.4	8.2	6.2	7.7	38	130
## 400	3.9	8.0	6.0	7.9	136	42000
## 401	4.3	7.9	6.1	7.7	38	73
## 402	3.9	8.1	6.0	7.5	33	140
## 403	6.2	8.0	6.7	8.0	53	2500
## 404	4.0	7.7	6.4	8.3	40	96
## 405	3.8	7.8	6.1	7.7	300	56000
## 406	7.1	8.8	6.5	8.0	60	80
## 407	3.0	8.0	6.5	7.9	80	59000
## 408	6.0	8.3	6.4	8.0	69	140
## 409	5.4	8.1	6.5	7.8	54	26700
## 410	3.8	8.4	7.4	8.1	200	1356
## 411	6.2	9.3	7.0	8.2	101	229
## 412	6.0	8.9	6.2	7.9	109	440
## 413	5.0	7.1	6.0	6.9	36	96
## 414	3.8	7.4	6.2	7.2	37	16700
## 415	1.1	6.8	6.0	6.6	47	407
## 416	3.0	7.3	6.5	7.7	37	33600
## 417	0.8	3.4	5.7	6.9	80	600
## 418	1.2	3.4	6.0	6.9	70	233
## 419	5.4	8.3	6.3	7.9	85	148
## 420	2.1	7.2	6.1	7.3	75	180
## 421	3.8	8.2	7.2	8.3	139	1082
## 422	6.3	8.1	6.7	7.8	48	101
## 423	6.3	7.4	7.1	7.4	100	120
## 424	5.8	6.8	6.4	6.8	90	118
## 425	5.8	7.5	6.5	6.9	36	66
## 426	6.1	8.4	6.1	6.9	102	150
## 427	5.8	7.3	6.9	7.8	62	45000
## 428	5.8	8.3	5.8	7.3	40	93
## 429	1.6	8.0	6.1	7.6	60	126
## 430	7.0	7.0	7.7	7.7	244	244
## 431	6.5	6.5	7.2	7.2	241	241
## 432	4.5	8.7	3.9	7.6	98	5951
## 433	5.2	9.3	6.9	7.6	79	751
## 434	4.6	9.9	7.1	8.2	148	1045
## 435	5.6	7.9	7.2	7.5	156	228
## 436	0.2	12.2	6.6	7.9	260	2216
## 437	0.4	9.7	6.5	8.3	116	2851
## 438	5.8	9.5	6.7	7.7	113	956
## 439	2.9	8.5	6.7	7.8	103	953

## 440	7.4	7.4	6.6	6.6	233	233
## 441	6.2	6.2	7.2	7.2	282	282
## 442	7.7	7.7	7.2	7.2	223	223
## 443	5.1	8.9	7.2	8.2	146	770
## 444	6.5	9.1	7.1	8.2	170	730
## 445	5.9	10.5	6.1	7.8	176	1035
## 446	6.1	9.9	7.0	7.5	151	289
## 447	1.5	13.2	6.7	8.0	192	1015
## 448	3.5	17.8	6.9	8.6	55	524
## 449	1.4	13.6	6.5	7.9	149	2192
## 450	4.7	10.1	6.8	8.2	308	1168
## 451	4.9	9.5	6.5	8.2	108	1050
## 452	3.6	8.9	6.9	8.1	161	1095
## 453	5.4	9.7	7.0	7.8	94	856
## 454	4.2	8.2	7.1	7.5	112	531
## 455	7.8	7.8	7.1	7.1	934	934
## 456	7.1	11.3	7.1	8.4	10	959
## 457	6.0	9.6	7.0	7.8	172	1211
## 458	6.3	9.3	6.9	8.4	164	910
## 459	6.4	12.2	7.0	8.7	113	485
## 460	6.2	6.2	7.4	7.4	198	198
## 461	2.4	10.3	7.0	8.2	124	949
## 462	5.2	7.6	6.8	7.7	108	921
## 463	6.0	7.4	7.0	7.4	67	90
## 464	6.2	7.4	2.4	4.1	610	1240
## 465	5.0	7.5	6.7	7.4	53	73
## 466	7.2	8.4	7.1	7.5	68	80
## 467	3.8	8.6	7.2	7.5	57	95
## 468	6.6	7.5	6.9	7.4	47	74
## 469	6.4	8.6	6.7	7.5	40	55
## 470	6.2	8.0	7.4	7.5	42	63
## 471	5.6	7.2	6.6	7.2	50	83
## 472	6.2	8.0	6.8	7.5	41	650
## 473	6.9	8.6	7.0	7.4	39	70
## 474	7.3	8.4	7.3	7.8	90	158
## 475	5.6	7.9	6.5	7.2	36	440
## 476	6.8	8.7	6.9	7.6	47	73
## 477	1.5	3.8	6.9	7.6	34	465
## 478	3.5	6.0	7.1	7.8	245	300
## 479	0.8	2.4	7.2	7.6	370	480
## 480	1.0	3.0	7.1	7.7	398	662
## 481	6.4	8.6	6.8	7.3	36	59
## 482	1.4	2.7	7.0	7.5	280	400
## 483	3.8	5.8	7.1	7.6	155	318
## 484	6.8	8.6	6.7	7.5	40	470
## 485	5.7	8.9	2.6	17.0	805	1450
## 486	7.1	9.0	7.0	7.8	110	374
## 487	6.8	8.3	4.9	6.2	42	75
## 488	0.5	8.0	7.3	7.5	72	138
## 489	7.1	8.0	7.4	7.6	76	100
## 490	6.2	10.5	7.4	8.4	260	2005
## 491	6.8	10.1	7.5	8.4	436	34020
## 492	6.3	7.0	6.7	8.4	171	2407
## 493	7.0	7.6	6.6	8.4	184	435

## 494	6.9	7.8	6.4	8.3	154	228
## 495	6.8	9.6	6.8	8.5	132	305
## 496	5.2	8.2	7.0	8.4	184	422
## 497	0.3	7.8	7.1	8.1	68	371
## 498	6.7	7.9	6.6	8.4	154	311
## 499	7.1	8.1	6.6	8.4	135	261
## 500	4.5	8.1	7.5	8.4	268	792
## 501	7.0	7.0	8.1	8.1	28200	28200
## 502	6.5	6.5	7.9	7.9	31400	31400
## 503	6.3	6.3	7.8	7.8	28000	28000
## 504	6.3	6.3	6.7	6.7	932	932
## 505	4.0	53.0	7.3	8.2	415	1013
## 506	BDL	3.4	7.2	7.9	607	2066
## 507	2.6	5.5	7.0	7.8	506	1349
## 508	BDL	4.0	7.1	8.0	607	2187
## 509	1.2	6.4	7.1	7.9	470	1223
## 510	2.1	4.3	7.2	8.1	490	1196
## 511	1.7	4.1	7.3	7.9	557	1307
## 512	1.0	3.7	7.1	8.3	756	1546
## 513	1.3	4.4	7.2	7.9	399	1320
## 514	BDL	24.0	6.6	8.0	418	1775
## 515	1.0	5.2	7.3	8.1	507	1710
## 516	1.2	5.6	7.1	8.1	498	1624
## 517	0.0	BDL	7.1	7.8	1290	2078
## 518	5.0	6.9	7.1	8.5	120	1260
## 519	5.5	5.5	7.5	7.5	540	540
## 520	4.3	4.3	7.4	7.4	490	490
## 521	4.4	6.6	7.7	8.2	230	580
## 522	3.8	5.2	8.1	8.3	640	780
## 523	2.9	6.8	8.3	8.5	340	1040
## 524	0.7	4.3	8.0	8.8	8000	15600
## 525	6.5	7.6	6.6	8.5	36	118
## 526	6.4	7.2	6.5	8.2	71	171
## 527	5.1	7.2	6.6	8.2	59	209
## 528	6.5	7.5	6.8	8.7	82	690
## 529	6.0	7.0	6.5	8.2	39	130
## 530	6.6	7.3	6.5	8.6	40	98
## 531	6.4	7.5	6.9	8.9	181	1138
## 532	6.3	6.3	7.5	7.5	774	774
## 533	6.3	7.5	7.0	8.7	163	482
## 534	6.4	7.5	6.5	8.5	40	112
## 535	6.1	7.5	7.4	8.7	109	442
## 536	4.2	6.9	6.7	8.2	70	240
## 537	4.0	7.3	6.7	8.2	71	245
## 538	5.1	7.0	7.0	8.9	78	161
## 539	5.0	6.9	6.2	8.0	89	169
## 540	5.7	8.4	6.0	8.0	77	157
## 541	5.1	6.8	6.2	7.6	84	157
## 542	4.5	6.8	6.1	7.9	94	169
## 543	4.3	7.2	6.3	8.2	79	169
## 544	5.2	7.2	6.0	7.4	88	173
## 545	5.1	7.4	6.1	7.7	107	179
## 546	6.2	7.9	7.0	9.0	92	164
## 547	6.9	8.9	6.6	7.9	82	161

## 548	5.1	7.1	6.1	8.0	81	177
## 549	6.8	8.8	6.3	8.5	92	164
## 550	6.2	7.9	5.8	7.1	71	140
## 551	6.5	8.2	6.7	8.6	88	148
## 552	5.3	7.5	6.6	8.5	89	167
## 553	6.3	8.9	7.0	8.8	91	159
## 554	6.1	7.8	6.2	7.9	72	155
## 555	5.6	7.6	6.1	8.2	88	159
## 556	6.1	7.9	6.9	8.5	81	149
## 557	6.5	8.6	5.8	7.2	73	161
## 558	4.6	6.9	6.2	8.0	85	149
## 559	5.1	7.0	6.1	7.9	83	166
## 560	5.4	8.0	6.3	8.1	79	157
## 561	5.8	7.8	6.8	8.6	83	159
## 562	6.6	10.0	7.0	8.4	82	159
## 563	5.1	7.1	5.9	7.6	83	148
## 564	5.8	6.8	7.1	8.4	228	60900
## 565	5.7	8.1	6.7	8.4	200	585
## 566	5.0	8.5	7.2	8.7	234	631
## 567	4.6	7.0	7.1	8.5	302	1090
## 568	6.5	8.1	6.7	8.3	254	384
## 569	5.4	8.2	6.7	8.6	240	462
## 570	6.4	7.2	7.1	8.0	50	100
## 571	5.0	7.0	7.2	8.4	5000	85900
## 572	4.1	5.8	7.7	7.9	28800	34100
## 573	5.5	8.6	7.0	8.2	225	375
## 574	6.0	8.0	7.4	8.6	197	652
## 575	4.8	6.8	6.8	8.1	52	130
## 576	5.8	6.8	7.4	8.1	73	131
## 577	-	-	7.3	7.5	158	169
## 578	2.0	9.4	7.3	8.6	48	193
## 579	-	-	6.8	7.7	283	468
## 580	5.0	8.6	6.7	7.4	34	70
## 581	3.3	5.6	6.6	7.7	20	43
## 582	4.3	6.5	6.4	7.8	72	212
## 583	5.2	6.6	7.5	8.0	72	106
## 584	3.8	10.0	6.6	7.6	58	85
## 585	4.3	9.2	7.4	8.1	66	209
## 586	3.7	7.0	6.9	7.7	75	192
## 587	6.7	9.9	6.8	8.0	25	41
## 588	0.8	1.4	6.9	7.5	781	2890
## 589	4.1	9.8	7.3	7.9	19	63
## 590	4.9	5.9	7.1	7.8	27	188
## 591	7.0	9.8	6.9	7.8	25	84
## 592	5.6	5.8	6.9	7.3	90	116
## 593	5.8	10.0	7.1	7.8	60	136
## 594	2.0	5.2	5.8	6.5	112	198
## 595	-	-	7.4	7.6	84	86
## 596	5.4	6.8	7.3	8.1	71	102
## 597	4.6	7.6	6.0	7.5	80	404
## 598	6.8	10.5	7.3	8.1	70	211
## 599	4.3	8.8	7.0	9.3	69	149
## 600	5.1	9.4	6.8	7.9	45	180
## 601	1.0	1.5	7.0	7.8	62	627

## 602	-	-	7.0	7.4	280	464
## 603	5.4	7.8	5.8	7.8	115	301
## 604	4.5	7.5	6.3	7.5	25	38
## 605	5.0	10.2	6.7	7.8	54	145
## 606	5.1	7.7	6.4	7.6	78	127
## 607	4.8	8.7	7.2	8.0	199	493
## 608	3.7	7.6	7.5	8.1	205	462
## 609	6.1	7.5	7.4	8.1	359	561
## 610	3.5	6.3	7.1	7.8	436	1160
## 611	1.6	1.6	7.8	7.8	1240	1240
## 612	-	-	7.1	7.1	1430	1430
## 613	-	-	7.9	7.9	1420	1420
## 614	3.4	3.4	7.7	7.7	1220	1220
## 615	10.0	10.0	7.9	7.9	458	458
## 616	4.2	7.2	7.3	7.9	136	2402
## 617	3.6	6.1	6.7	7.9	228	47020
## 618	4.3	6.2	7.6	8.3	13330	52350
## 619	5.1	6.7	7.4	8.4	2030	55740
## 620	4.1	6.9	7.4	8.6	199	572
## 621	4.5	6.8	7.2	8.3	210	5843
## 622	5.4	7.3	7.2	9.3	216	600
## 623	6.0	7.3	7.5	8.9	198	3584
## 624	3.4	6.2	7.3	7.9	237	47400
## 625	6.4	9.3	6.1	7.8	45	90
## 626	5.7	11.0	6.2	7.5	14	2851
## 627	7.0	8.3	6.7	7.4	56	460
## 628	4.0	8.2	6.1	8.1	27	48300
## 629	6.1	8.0	6.4	7.9	18	6725
## 630	6.1	7.8	6.1	7.8	51	1132
## 631	6.9	8.6	6.3	7.4	37	1013
## 632	5.1	8.3	6.4	8.0	23	47750
## 633	6.9	8.6	6.2	8.5	41	126
## 634	4.9	7.2	6.6	8.1	283	48800
## 635	4.7	7.2	6.5	7.5	121	46370
## 636	4.7	7.0	6.6	7.9	210	45470
## 637	2.6	9.3	6.2	7.5	87	11180
## 638	5.1	7.3	6.3	7.7	68	276
## 639	5.1	7.2	6.8	8.0	4810	52590
## 640	BDL	4.1	6.3	7.9	97	478
## 641	4.0	7.7	6.5	7.9	242	34780
## 642	6.2	7.2	5.9	7.1	32	269
## 643	5.0	6.4	6.1	8.0	7580	53160
## 644	5.0	6.3	6.1	7.9	7600	53220
## 645	6.6	8.4	6.2	7.8	42	5789
## 646	7.3	8.7	6.3	7.6	42	66
## 647	4.3	7.4	6.3	8.1	37	48600
## 648	5.0	7.7	6.4	7.6	84	23600
## 649	4.5	7.7	6.2	7.7	211	38490
## 650	4.6	7.4	6.0	7.8	891	51740
## 651	6.1	7.2	7.2	8.1	245	51530
## 652	BDL	4.4	7.2	8.1	865	2720
## 653	BDL	4.4	7.1	7.8	876	3230
## 654	5.3	7.0	7.1	8.2	290	1969
## 655	BDL	6.7	7.4	8.8	717	11920

## 656	6.4	7.5	7.1	7.9	218	299
## 657	2.4	6.9	7.4	8.0	92	38138
## 658	BDL	1.8	7.1	7.9	530	1673
## 659	6.7	6.7	8.3	8.5	168	381
## 660	6.0	7.3	7.1	8.0	255	73685
## 661	3.8	5.4	7.2	8.0	174	14870
## 662	BDL	6.7	7.2	7.4	633	4238
## 663	5.2	7.2	7.1	7.9	668	46892
## 664	6.1	7.2	7.1	8.2	381	20490
## 665	7.1	7.1	7.5	7.5	1313	1313
## 666	5.0	6.5	7.1	8.4	757	46400
## 667	BDL	3.1	7.0	7.8	305	1554
## 668	6.3	7.7	7.5	8.4	248	637
## 669	1.7	6.8	7.6	10.3	337	17900
## 670	2.8	4.9	8.2	9.3	800	16900
## 671	2.5	6.8	2.8	8.0	352	3570
## 672	4.6	8.2	7.0	8.4	324	667
## 673	1.9	4.4	8.0	10.4	782	21600
## 674	7.8	9.4	8.1	8.6	381	418
## 675	7.6	11.0	7.4	8.3	211	623
## 676	5.2	9.5	7.4	8.4	201	3550
## 677	7.1	8.9	7.2	8.1	57	236
## 678	7.2	8.4	7.0	8.1	83	254
## 679	5.7	8.4	6.4	8.3	368	4690
## 680	6.0	9.2	7.0	8.2	375	587
## 681	6.4	9.6	7.1	8.3	378	592
## 682	5.6	9.9	7.0	8.0	369	4640
## 683	3.0	7.5	6.5	8.2	386	1033
## 684	5.8	9.0	6.9	8.2	362	6170
## 685	8.2	8.6	7.1	8.1	109	1230
## 686	8.1	8.6	7.0	8.5	7	246
## 687	8.5	11.6	7.6	8.2	62	122
## 688	8.2	8.2	8.4	8.4	188	188
## 689	7.4	8.2	7.6	8.3	96	136
## 690	8.5	8.5	8.1	8.1	119	119
## 691	4.6	7.2	6.8	7.6	202	282
## 692	-	-	6.0	6.6	-	-
## 693	6.8	6.8	6.5	7.5	-	-
## 694	7.2	7.5	7.2	7.6	-	-
## 695	7.0	7.6	7.2	7.5	-	-
## 696	3.5	6.2	7.3	7.6	235	332
## 697	7.2	7.7	7.4	7.5	202	238
## 698	8.3	8.5	7.3	7.5	-	-
## 699	6.5	7.8	8.0	8.2	338	740
## 700	6.4	7.4	8.0	8.3	450	585
## 701	7.2	8.2	6.9	7.6	50	110
## 702	7.6	8.1	7.0	7.4	47	108
## 703	6.9	7.8	6.9	7.8	59	119
## 704	6.2	7.0	6.8	7.3	134	180
## 705	6.0	6.7	6.8	7.4	136	178
## 706	6.0	7.0	6.9	7.4	130	176
## 707	6.0	7.0	6.8	7.3	138	172
## 708	5.3	6.5	6.5	7.3	114	134
## 709	5.1	6.3	6.0	7.1	116	132

## 710	6.0	7.4	6.0	7.3	102	128
## 711	5.4	7.2	6.5	7.4	102	124
## 712	4.2	8.0	6.4	7.4	48	135
## 713	4.6	8.8	6.1	7.4	56	22000
## 714	1.5	6.4	5.9	7.7	100	182
## 715	5.3	7.2	6.0	8.2	96	360
## 716	5.7	7.1	6.5	8.2	160	508
## 717	3.6	8.6	6.0	7.4	40	375
## 718	6.1	8.7	6.2	7.9	26	128
## 719	5.1	9.2	6.1	7.7	34	42100
## 720	6.0	8.2	6.1	8.1	28	160
## 721	6.3	8.7	6.3	8.1	39	120
## 722	1.4	7.1	6.4	7.1	149	6500
## 723	4.1	7.9	6.4	7.9	218	33250
## 724	6.0	6.9	7.0	7.6	118	130
## 725	5.9	6.8	6.8	7.8	9140	9890
## 726	4.0	8.0	6.0	7.9	65	158
## 727	4.6	8.3	6.5	7.6	60	140
## 728	0.8	4.8	6.5	7.3	94	210
## 729	0.8	8.4	6.5	7.6	88	160
## 730	6.2	7.1	6.5	7.4	104	118
## 731	2.1	5.8	7.0	7.9	126	52300
## 732	2.1	6.9	6.6	8.1	133	1296
## 733	5.1	8.0	6.5	7.1	100	190
## 734	4.1	8.7	6.4	8.1	54	105
## 735	0.1	2.8	6.4	7.2	405	935
## 736	5.0	7.7	6.0	7.3	70	186
## 737	5.0	7.5	6.0	7.5	65	121
## 738	3.6	6.3	5.8	7.3	42	20700
## 739	4.9	7.1	6.0	7.8	88	195
## 740	5.4	7.3	6.2	8.0	300	1524
## 741	4.9	8.2	6.2	7.5	64	41800
## 742	4.8	8.0	6.8	7.8	32	125
## 743	6.8	9.9	6.2	7.7	24	44
## 744	5.7	8.6	6.4	7.7	33	102
## 745	4.1	7.4	6.1	7.8	102	189
## 746	5.4	8.1	6.9	7.9	80	330
## 747	5.8	8.0	6.8	7.6	84	304
## 748	4.2	7.1	6.5	7.6	66	199
## 749	3.0	8.5	6.2	8.8	54	228
## 750	6.0	7.7	6.8	7.9	44	103
## 751	4.1	7.8	6.2	7.9	199	49800
## 752	5.7	7.8	6.9	7.9	54	86
## 753	3.0	8.2	6.3	8.1	75	40000
## 754	5.9	8.2	6.2	8.1	53	820
## 755	3.6	7.0	6.0	7.3	80	157
## 756	4.1	7.1	6.1	7.6	92	147
## 757	6.0	7.1	6.5	7.2	120	132
## 758	3.1	5.8	6.1	7.0	68	100
## 759	2.1	6.6	6.3	6.9	56	88
## 760	6.1	8.0	6.4	8.1	46	90
## 761	5.2	7.3	6.1	7.9	59	115
## 762	5.3	8.0	6.1	6.8	43	72
## 763	4.4	9.6	6.6	7.7	90	33000

## 764	4.0	8.2	6.6	7.6	32	56
## 765	4.8	7.7	4.1	7.3	38	78
## 766	4.5	7.9	6.8	7.7	35	92
## 767	4.7	9.1	6.5	7.2	44	29000
## 768	0.9	7.4	6.3	7.3	41	135
## 769	5.3	7.2	6.7	7.7	49	9600
## 770	5.3	8.8	6.5	7.5	30	64
## 771	4.4	7.1	6.3	7.3	39	1100
## 772	4.9	9.1	5.2	8.2	37	78
## 773	5.0	7.5	5.8	7.7	85	182
## 774	3.2	8.3	6.1	7.7	131	53500
## 775	7.0	9.7	6.1	7.0	39	66
## 776	1.3	8.0	6.2	7.4	24	122
## 777	2.1	6.7	6.0	7.5	74	39650
## 778	4.4	9.1	6.1	7.9	62	46500
## 779	4.5	8.2	6.9	7.6	66	42600
## 780	3.1	7.4	6.2	7.8	75	223
## 781	6.7	7.3	7.8	8.6	254	414
## 782	5.6	7.4	7.7	8.3	289	602
## 783	6.5	7.7	7.3	7.8	258	520
## 784	6.2	7.8	7.4	8.3	170	3420
## 785	5.4	6.7	7.1	8.3	310	682
## 786	6.5	7.4	7.3	8.4	167	288
## 787	6.2	7.2	7.8	8.7	300	521
## 788	6.0	7.5	7.3	7.9	198	593
## 789	6.2	7.6	7.4	7.9	267	541
## 790	5.0	7.5	6.7	8.9	71	810
## 791	4.4	7.5	6.9	8.4	79	50920
## 792	6.0	7.4	6.5	8.7	67	1970
## 793	6.8	7.6	6.6	8.5	66	385
## 794	5.0	7.5	6.7	8.0	71	3572
## 795	3.8	6.2	7.2	8.0	374	819
## 796	6.4	7.6	6.9	9.0	65	137
## 797	4.4	7.2	7.0	8.0	93	59500
## 798	3.5	7.4	6.5	8.1	85	66200
## 799	4.7	7.6	6.8	8.8	67	590
## 800	BDL	7.3	6.1	7.8	129	58970
## 801	4.0	7.0	6.5	8.1	64	55900
## 802	6.2	7.2	6.6	7.7	80	233
## 803	4.3	7.2	6.6	8.2	83	1010
## 804	3.4	7.2	6.6	8.5	127	23800
## 805	6.4	7.4	6.5	7.9	74	997
## 806	6.1	7.3	6.5	7.7	73	277
## 807	6.4	7.4	6.5	7.8	73	176
## 808	3.9	7.2	6.5	7.8	64	1002
## 809	5.4	7.4	6.8	8.5	87	15710
## 810	3.8	7.0	6.5	7.8	47	55780
## 811	4.1	7.1	6.5	8.2	49	54950
## 812	4.0	7.1	6.7	8.0	49	55750
## 813	BDL	6.9	6.5	7.7	51	29790
## 814	4.0	6.9	7.0	7.7	50	58790
## 815	5.8	7.6	6.5	8.5	121	212
## 816	6.0	7.5	6.7	8.5	79	271
## 817	4.3	7.4	6.8	8.4	97	58300

## 818	6.9	7.4	7.3	8.0	74	288
## 819	6.4	7.4	6.7	8.0	97	3285
## 820	6.0	7.4	6.6	7.9	64	958
## 821	6.3	7.5	6.6	8.0	77	188
## 822	7.0	7.3	7.2	8.0	82	277
## 823	5.5	7.0	6.7	8.1	60	397
## 824	4.0	6.9	6.9	7.8	54	12900
## 825	5.2	7.0	7.3	7.8	59	10410
## 826	3.1	8.1	6.9	7.7	105	408
## 827	6.9	7.9	7.1	7.7	100	516
## 828	7.3	8.2	7.0	7.8	125	4225
## 829	3.1	7.1	6.8	7.6	120	445
## 830	5.3	7.6	6.9	7.8	190	460
## 831	4.1	7.7	6.3	7.4	110	420
## 832	3.9	7.8	7.0	7.9	90	482
## 833	6.7	6.7	7.6	7.6	234	234
## 834	6.8	8.1	7.0	7.7	130	482
## 835	7.1	8.3	7.2	7.9	94	411
## 836	7.0	8.1	7.0	7.6	115	390
## 837	5.3	7.8	7.1	7.6	150	458
## 838	6.9	8.1	6.9	7.6	65	407
## 839	4.4	7.9	7.0	7.7	91	340
## 840	6.6	7.9	7.1	7.8	65	460
## 841	6.8	7.6	7.0	7.6	135	452
## 842	6.4	7.8	7.0	7.5	150	470
## 843	7.2	8.1	7.1	7.6	150	401
## 844	7.2	8.2	7.1	7.4	190	476
## 845	6.3	7.8	7.1	7.8	130	439
## 846	3.7	6.2	6.0	7.9	270	1410
## 847	3.8	5.6	6.2	7.9	270	1220
## 848	4.6	5.9	6.0	7.8	288	643
## 849	6.8	8.5	7.1	7.5	70	224
## 850	4.1	6.8	5.8	7.9	160	624
## 851	4.8	7.8	6.9	7.4	130	516
## 852	7.1	7.9	7.1	7.8	29	389
## 853	6.9	7.9	6.9	7.6	170	550
## 854	7.0	8.3	7.0	7.6	170	305
## 855	7.2	8.5	7.1	7.9	122	360
## 856	7.1	8.4	7.0	7.4	127	387
## 857	7.2	8.3	6.9	7.6	120	420
## 858	6.8	7.5	7.0	7.8	62	144
## 859	6.3	7.5	6.8	7.8	71	150
## 860	7.0	8.0	7.1	7.8	64	87
## 861	6.3	7.7	7.1	7.5	68	110
## 862	6.0	7.6	2.7	4.4	190	700
## 863	5.0	7.2	2.3	3.3	1000	1480
## 864	7.4	8.0	6.7	7.3	142	430
## 865	6.1	7.4	6.6	7.1	59	72
## 866	6.0	7.7	5.0	6.6	40	72
## 867	7.0	7.7	7.1	7.6	67	94
## 868	6.8	7.7	6.3	7.4	62	89
## 869	7.0	7.7	6.9	7.5	72	89
## 870	6.0	7.6	6.8	7.2	50	130
## 871	6.4	7.6	6.8	7.3	40	72

## 872	7.0	7.6	6.7	7.3	30	44
## 873	6.9	8.0	7.6	7.9	108	130
## 874	6.5	7.4	7.0	7.3	48	66
## 875	6.6	7.9	6.8	7.5	41	58
## 876	7.0	7.6	7.3	7.8	44	60
## 877	0.8	3.8	6.8	7.6	215	737
## 878	2.5	8.2	6.8	7.5	225	380
## 879	0.5	3.4	7.3	7.9	285	450
## 880	0.8	3.4	6.9	8.1	427	550
## 881	6.4	7.9	6.7	7.2	39	55
## 882	7.0	8.2	7.0	7.6	50	75
## 883	6.9	7.8	6.7	7.3	37	58
## 884	0.5	3.5	6.8	7.6	170	465
## 885	3.0	6.0	6.7	7.3	15	458
## 886	5.4	7.5	6.9	7.4	67	84
## 887	6.0	7.6	6.9	8.0	25	84
## 888	6.6	6.6	7.7	7.7	7190	7190
## 889	7.0	7.0	7.7	7.7	9110	9110
## 890	6.6	6.6	7.5	7.5	12940	12940
## 891	6.3	6.3	7.4	7.4	2940	2940
## 892	4.3	6.9	7.1	8.1	288	630
## 893	1.8	3.8	7.1	8.1	472	1248
## 894	2.7	4.7	7.2	7.8	435	898
## 895	1.8	4.3	7.3	8.0	620	1452
## 896	1.9	4.9	7.1	7.8	417	952
## 897	2.7	4.4	7.3	7.8	328	998
## 898	2.3	4.0	7.2	7.9	532	1200
## 899	2.6	3.6	7.3	7.9	451	1240
## 900	3.1	4.4	7.2	7.8	323	980
## 901	2.2	4.2	7.1	8.0	510	1330
## 902	BDL	5.9	7.5	7.9	394	1410
## 903	BDL	6.5	7.4	7.9	313	1390
## 904	1.6	4.3	7.0	8.1	651	1840
## 905	BDL	3.6	7.0	8.0	750	2040
## 906	BDL	BDL	7.0	8.0	731	1227
## 907	-	-	7.8	8.4	140	570
## 908	-	-	7.4	8.6	300	610
## 909	-	-	7.8	8.7	270	1810
## 910	5.3	6.7	7.3	7.3	920	1389
## 911	6.2	8.0	6.0	7.1	26	307
## 912	6.6	7.6	5.8	6.8	36	692
## 913	4.3	7.4	6.1	7.1	51	1132
## 914	5.9	7.9	6.3	7.4	90	274
## 915	5.0	7.9	5.7	6.8	29	481
## 916	6.4	8.2	5.8	7.0	28	471
## 917	3.0	7.5	6.2	7.7	135	2540
## 918	4.3	7.8	6.5	7.3	255	565
## 919	6.7	8.0	5.8	6.9	32	491
## 920	4.0	7.0	6.3	7.5	160	1044
## 921	2.5	7.4	6.2	6.8	55	184
## 922	2.9	7.1	6.2	7.0	56	253
## 923	6.0	6.0	7.0	7.0	286	286
## 924	6.5	7.9	6.6	7.4	114	189
## 925	6.2	7.6	6.9	7.4	106	177

## 926	6.2	7.5	6.9	7.8	113	174
## 927	5.7	7.5	7.0	7.7	102	177
## 928	6.3	7.6	7.0	7.4	119	193
## 929	6.5	7.9	6.8	7.7	111	167
## 930	6.5	7.9	6.7	7.4	117	181
## 931	6.0	7.4	6.1	7.7	111	191
## 932	5.8	7.7	6.7	7.7	118	212
## 933	6.1	7.9	7.0	7.7	106	183
## 934	6.6	7.7	6.6	7.4	111	169
## 935	6.5	8.0	6.2	7.5	109	162
## 936	6.1	7.6	6.6	7.3	111	163
## 937	5.5	7.8	6.6	7.7	112	189
## 938	6.2	7.6	6.6	7.7	121	233
## 939	6.1	7.9	6.6	7.5	109	174
## 940	6.2	7.9	6.8	7.7	112	198
## 941	6.3	8.1	6.7	7.6	105	175
## 942	5.9	7.8	6.8	7.3	112	173
## 943	6.2	7.3	6.6	7.3	110	172
## 944	6.5	7.9	6.3	7.3	110	179
## 945	6.7	7.7	6.7	7.6	111	188
## 946	6.3	7.6	6.6	7.3	106	175
## 947	6.5	7.9	7.0	7.4	109	167
## 948	6.2	7.8	6.9	7.6	105	177
## 949	5.8	7.8	6.9	7.5	110	179
## 950	Min	Max	Min	Max	Min	Max
## 951	5.0	8.0	7.4	8.4	212	440
## 952	5.8	8.5	6.9	8.3	228	321
## 953	6.0	8.5	7.2	8.3	274	522
## 954	5.3	8.0	7.1	8.7	324	790
## 955	6.0	8.2	6.9	8.2	248	369
## 956	5.6	8.0	7.4	8.3	212	408
## 957	3.7	5.2	7.5	8.1	28400	47930
## 958	5.6	6.8	6.4	7.6	42	92
## 959	5.7	5.7	8.1	8.1	732	732
## 960	5.2	6.6	7.0	8.0	98	130
## 961	5.0	6.8	6.6	7.5	74	96
## 962	5.5	6.4	7.0	7.7	96	127
## 963	5.8	6.7	7.1	7.8	71	98
## 964	5.1	6.8	6.9	7.8	108	205
## 965	6.0	7.5	6.8	7.6	115	141
## 966	5.1	7.8	6.3	7.3	187	297
## 967	4.9	7.6	6.8	7.5	26	61
## 968	5.0	6.5	7.0	7.8	94	230
## 969	6.5	8.8	6.9	7.8	142	180
## 970	0.3	0.3	7.2	7.6	108	176
## 971	4.2	9.4	7.0	7.9	98	119
## 972	5.6	9.8	7.3	8.0	126	150
## 973	5.2	8.0	7.0	7.8	165	196
## 974	5.7	8.8	7.0	8.1	158	180
## 975	5.2	9.8	7.3	7.7	39	69
## 976	4.9	5.9	7.0	7.7	40	79
## 977	5.2	8.8	7.2	7.7	23	82
## 978	5.3	5.8	7.1	7.5	90	125
## 979	6.5	9.8	7.0	7.7	47	96

## 980	5.6	9.5	7.1	7.6	46	94
## 981	5.6	9.4	7.0	7.9	54	133
## 982	6.6	9.8	7.4	8.0	203	352
## 983	5.5	8.5	7.5	8.3	169	668
## 984	1.1	7.5	7.3	7.6	209	771
## 985	4.8	8.5	7.2	7.9	130	354
## 986	6.1	7.2	7.3	7.9	145	274
## 987	4.3	6.3	7.1	7.7	217	7985
## 988	4.5	4.9	7.3	8.0	7050	28710
## 989	4.3	6.0	7.1	8.0	1210	32820
## 990	4.4	6.8	7.2	8.3	228	385
## 991	5.9	7.3	7.3	8.4	186	336
## 992	5.1	7.2	7.5	8.3	224	367
## 993	5.9	7.2	7.5	8.4	168	296
## 994	3.5	6.1	6.9	8.1	325	4250
## 995	1.4	5.3	6.8	7.3	-	-
## 996	4.5	7.9	7.0	7.7	258	58470
## 997	5.3	7.5	6.7	7.9	359	59130
## 998	4.0	7.6	6.2	7.5	82	27560
## 999	4.0	7.3	6.1	7.2	135	47660
## 1000	6.4	8.0	6.5	7.2	15	101
## 1001	6.4	8.4	5.8	7.0	44	94
## 1002	5.1	8.0	5.4	7.8	61	297
## 1003	6.5	7.6	6.3	7.7	12	163
## 1004	6.6	7.8	6.3	8.1	46	122
## 1005	5.7	7.2	5.6	6.9	50	127
## 1006	6.5	8.0	6.1	7.7	12	155
## 1007	4.5	7.0	6.3	7.4	40	250
## 1008	6.6	8.0	6.1	7.6	15	209
## 1009	4.3	6.5	6.8	7.9	23320	65600
## 1010	2.0	7.1	6.2	7.3	30	656
## 1011	6.1	8.5	6.5	7.3	16	7078
## 1012	5.8	9.3	5.9	7.7	72	938
## 1013	3.8	7.4	6.3	7.2	196	37180
## 1014	4.0	7.5	6.4	7.8	83	60000
## 1015	0.3	5.6	6.6	7.2	89	321
## 1016	3.8	5.5	6.2	7.4	175	36810
## 1017	4.5	7.5	6.8	7.7	722	62100
## 1018	4.7	7.2	6.9	7.7	830	63430
## 1019	6.8	7.7	6.0	8.4	85	141
## 1020	6.2	8.1	6.5	8.5	60	110
## 1021	5.3	7.8	6.7	7.7	87	56140
## 1022	6.0	7.4	7.0	7.8	536	50290
## 1023	5.2	6.3	7.2	8.0	469	2612
## 1024	6.6	7.2	7.0	8.2	160	298
## 1025	4.8	7.5	4.9	7.9	265	9286
## 1026	0.3	3.6	7.2	8.2	1176	2380
## 1027	5.8	7.0	7.5	8.3	858	45790
## 1028	0.6	7.2	7.3	8.4	642	9787
## 1029	4.8	7.0	7.1	7.9	5	2296
## 1030	4.9	7.3	7.1	7.9	519	2086
## 1031	5.0	7.2	7.1	7.7	387	28630
## 1032	5.3	7.4	7.2	7.9	1306	52190
## 1033	0.3	4.1	7.1	7.8	592	1448

## 1034	0.3	5.2	7.3	8.1	892	2070
## 1035	0.6	5.9	7.2	8.6	590	2163
## 1036	5.0	7.2	7.6	8.4	706	13950
## 1037	0.3	2.5	6.7	8.0	635	1843
## 1038	4.8	8.3	7.6	8.2	5	862
## 1039	7.7	8.1	7.4	8.6	260	260
## 1040	2.1	6.8	7.9	8.8	1356	3800
## 1041	2.1	7.4	7.6	8.7	1468	3450
## 1042	1.8	7.9	6.3	8.1	430	1804
## 1043	6.8	9.2	7.0	8.8	382	938
## 1044	0.9	6.4	8.0	9.1	1318	3830
## 1045	4.2	5.1	7.5	7.8	135	904
## 1046	7.6	8.3	7.8	8.4	398	501
## 1047	3.8	3.8	7.7	7.7	1014	1014
## 1048	3.5	6.5	7.9	8.1	877	970
## 1049	3.2	3.2	8.6	8.6	651	651
## 1050	7.8	9.1	6.6	7.6	409	833
## 1051	7.6	9.2	7.1	7.9	303	564
## 1052	7.7	9.0	7.1	7.9	356	579
## 1053	6.1	9.1	7.1	7.7	303	774
## 1054	7.7	9.0	6.9	7.8	285	3210
## 1055	8.7	9.4	7.2	8.5	76	230
## 1056	8.1	9.0	7.1	8.4	82	263
## 1057	7.2	8.9	7.6	8.0	102	181
## 1058	7.4	9.0	7.5	8.1	115	154
## 1059	6.7	9.6	6.9	7.9	248	453
## 1060	6.7	9.7	6.8	7.9	251	622
## 1061	7.1	8.2	7.0	8.5	279	1862
## 1062	8.5	11.4	6.5	8.2	59	845
## 1063	8.5	10.1	7.3	8.1	65	257
## 1064	7.5	13.8	7.2	8.1	67	1179
## 1065	8.0	9.2	7.9	8.8	60	415
## 1066	7.5	9.1	7.2	8.5	265	565
## 1067	7.8	11.0	7.1	8.1	73	122
## 1068	7.6	9.5	7.2	8.1	59	338
## 1069	6.9	9.4	7.2	7.9	66	368
## 1070	5.8	7.8	6.4	9.0	256	636
## 1071	7.2	9.8	7.9	8.6	377	452
## 1072	8.2	10.1	7.8	8.5	310	375
## 1073	4.8	8.6	7.6	8.4	230	310
## 1074	7.4	7.7	7.2	7.6	-	-
## 1075	7.0	7.2	7.2	7.4	-	-
## 1076	6.8	7.2	7.4	7.5	166	202
## 1077	4.2	6.0	7.3	7.7	198	268
## 1078	7.4	7.5	7.4	7.4	202	205
## 1079	2.0	2.0	5.7	7.2	-	-
## 1080	2.8	10.2	6.5	7.1	-	-
## 1081	7.2	7.7	6.9	7.4	52	140
## 1082	6.7	7.5	6.7	7.4	122	166
## 1083	6.8	7.6	6.6	7.4	119	168
## 1084	6.9	7.8	6.7	7.3	46	131
## 1085	6.4	7.6	6.7	7.5	122	160
## 1086	5.9	7.4	6.9	7.4	124	156
## 1087	7.2	7.8	7.1	7.7	81	168

## 1088	7.1	7.9	6.9	7.8	63	150
## 1089	7.2	7.9	6.8	7.5	58	103
## 1090	7.6	8.1	6.8	7.9	51	121
## 1091	7.3	7.9	6.7	7.9	46	81
## 1092	7.1	7.7	6.7	7.5	54	119
## 1093	6.4	7.6	7.7	8.7	393	600
## 1094	6.4	7.8	7.7	8.7	338	560
## 1095	1.3	9.7	6.7	8.1	40	32400
## 1096	6.3	8.2	6.4	7.2	28	50
## 1097	6.7	8.6	6.0	7.7	32	83
## 1098	6.9	8.2	6.5	8.1	45	25120
## 1099	6.5	7.2	6.9	8.2	136	169
## 1100	6.4	8.1	6.6	8.2	45	68
## 1101	5.2	8.2	6.3	7.4	41	200
## 1102	0.2	2.5	6.1	7.3	289	618
## 1103	6.0	7.4	6.3	7.5	39	72
## 1104	6.7	8.4	6.0	7.4	38	60
## 1105	6.5	7.7	6.6	8.1	12	104
## 1106	5.9	8.8	6.6	7.8	98	292
## 1107	3.3	6.7	6.1	7.0	58	103
## 1108	5.8	7.2	6.7	7.9	128	154
## 1109	6.1	8.2	6.5	7.4	102	280
## 1110	5.9	6.9	6.4	7.6	66	96
## 1111	6.0	6.9	6.4	8.1	148	179
## 1112	4.7	7.1	6.2	7.4	56	120
## 1113	6.2	7.2	6.7	8.2	137	177
## 1114	3.2	6.3	6.2	6.9	46	92
## 1115	3.9	8.4	6.3	7.7	66	202
## 1116	7.4	8.3	6.5	7.8	22	74
## 1117	6.8	8.0	6.5	7.9	30	65
## 1118	4.5	8.0	6.3	7.9	314	46100
## 1119	5.2	7.9	6.5	8.1	89	55100
## 1120	6.4	8.4	6.1	8.2	52	123
## 1121	5.5	9.5	6.7	7.7	59	85
## 1122	0.7	8.0	5.4	7.2	110	9800
## 1123	4.4	6.8	6.2	7.4	56	130
## 1124	6.1	7.4	6.3	7.4	84	140
## 1125	5.3	6.8	6.1	7.5	71	153
## 1126	6.0	7.0	6.7	8.0	9240	9980
## 1127	6.0	6.9	6.5	8.1	153	186
## 1128	5.5	8.0	5.6	7.3	56	124
## 1129	3.4	8.2	6.0	7.2	96	180
## 1130	5.1	8.0	6.0	7.5	89	170
## 1131	3.0	6.8	4.1	7.6	82	29050
## 1132	5.5	9.6	6.0	7.5	67	190
## 1133	1.9	7.4	6.5	8.0	152	50100
## 1134	5.7	8.0	6.4	7.8	158	43600
## 1135	5.7	7.7	6.2	7.6	51	102
## 1136	5.5	7.7	6.4	7.4	43	91
## 1137	6.5	8.0	6.0	7.9	35	89
## 1138	4.1	5.7	6.3	7.9	409	47300
## 1139	5.2	7.9	6.2	7.8	37	152
## 1140	5.5	7.9	6.1	7.3	45	93
## 1141	5.2	7.9	6.0	7.7	55	7885

## 1142	5.4	8.1	6.2	7.8	52	98
## 1143	4.9	7.5	6.2	7.4	175	60900
## 1144	5.3	8.6	6.5	7.9	57	102
## 1145	5.9	8.2	6.5	7.8	98	58200
## 1146	6.4	8.2	6.8	7.5	70	151
## 1147	6.2	7.9	6.1	7.7	44	98
## 1148	5.2	6.8	7.0	8.0	271	1278
## 1149	6.3	7.9	6.9	8.0	89	317
## 1150	6.1	7.8	6.9	8.1	142	429
## 1151	6.5	8.2	6.1	7.1	34	141
## 1152	4.3	9.0	6.0	8.0	67	26000
## 1153	3.4	8.1	6.2	7.1	45	134
## 1154	6.1	8.2	6.6	7.9	33	8183
## 1155	0.6	7.0	6.0	7.1	90	6064
## 1156	1.5	7.6	6.0	7.2	60	537
## 1157	5.4	7.9	6.6	7.1	90	160
## 1158	4.9	6.9	6.6	7.3	55	149
## 1159	5.4	7.0	6.7	7.9	165	853
## 1160	7.1	8.0	6.7	7.8	45	92
## 1161	6.7	7.3	6.8	7.9	135	181
## 1162	6.7	7.1	6.6	8.1	137	186
## 1163	6.7	8.2	6.3	7.2	35	50
## 1164	5.6	7.2	6.2	7.6	55	80
## 1165	6.0	7.8	6.7	8.0	512	48100
## 1166	6.0	8.3	6.8	7.9	42	79
## 1167	1.9	7.7	6.5	7.6	52	103
## 1168	6.5	7.2	7.3	7.9	168	297
## 1169	6.8	8.0	7.5	8.6	288	652
## 1170	6.8	7.6	7.4	7.9	360	692
## 1171	7.0	7.5	7.0	8.3	246	718
## 1172	6.2	7.8	7.2	8.3	150	1738
## 1173	6.8	7.6	7.2	8.3	170	398
## 1174	1.0	7.6	7.8	8.5	249	50403
## 1175	6.7	7.7	7.6	8.3	240	489
## 1176	6.1	7.8	7.4	8.2	290	572
## 1177	5.6	7.3	7.2	8.3	160	348
## 1178	4.3	7.1	7.2	8.3	160	356
## 1179	3.1	7.5	7.0	8.3	111	3451
## 1180	6.3	7.7	7.1	8.3	129	276
## 1181	6.3	7.3	7.2	8.2	84	376
## 1182	4.9	7.3	6.7	8.1	69	4155
## 1183	6.3	7.6	6.8	8.3	77	192
## 1184	6.3	7.4	6.8	8.2	73	181
## 1185	6.1	7.3	6.8	8.2	70	1443
## 1186	6.1	7.4	7.0	8.2	100	392
## 1187	6.4	7.4	6.8	7.9	45	109
## 1188	0.3	6.8	6.9	7.9	303	49570
## 1189	4.2	6.9	6.7	8.2	80	60060
## 1190	5.4	7.4	7.2	7.8	103	32440
## 1191	6.2	7.7	6.8	8.1	68	195
## 1192	6.2	7.6	7.1	8.0	62	111
## 1193	6.5	7.7	7.0	8.1	62	119
## 1194	5.9	6.0	7.5	8.0	445	663
## 1195	4.3	7.0	6.7	8.1	108	63280

## 1196	6.3	7.7	6.7	8.1	67	179
## 1197	6.0	7.3	7.1	8.1	76	155
## 1198	4.2	7.3	6.7	7.5	71	57950
## 1199	6.0	7.6	6.9	8.2	70	1096
## 1200	6.0	7.3	7.0	8.2	67	4142
## 1201	5.8	7.4	7.0	8.2	68	4163
## 1202	5.8	7.4	6.9	8.1	69	26000
## 1203	6.0	7.4	7.0	8.2	109	178
## 1204	6.2	7.5	7.3	8.1	96	248
## 1205	5.9	7.5	6.9	8.0	90	293
## 1206	6.0	7.6	7.0	8.1	88	23650
## 1207	6.0	7.4	7.1	8.0	89	26810
## 1208	6.2	7.7	6.9	8.1	120	156
## 1209	6.5	7.6	7.2	8.1	134	247
## 1210	6.1	7.5	7.2	8.2	132	305
## 1211	7.0	7.2	7.3	8.2	118	309
## 1212	7.0	7.1	7.4	8.2	162	240
## 1213	6.4	7.2	6.7	7.9	46	162
## 1214	6.4	7.4	7.1	7.9	46	154
## 1215	6.1	7.5	7.1	8.0	70	336
## 1216	7.2	8.1	7.4	7.7	110	336
## 1217	6.5	7.8	7.2	7.9	150	420
## 1218	7.0	7.7	7.2	7.6	110	338
## 1219	3.1	7.2	6.5	7.6	320	624
## 1220	3.3	6.4	6.7	7.3	350	995
## 1221	6.2	7.6	7.2	7.8	125	250
## 1222	6.4	7.5	7.3	7.7	120	336
## 1223	6.8	7.7	7.0	7.5	150	418
## 1224	2.6	7.5	7.1	7.8	150	560
## 1225	6.4	7.7	7.2	7.8	135	365
## 1226	7.1	7.7	7.2	8.0	150	382
## 1227	6.5	7.4	7.1	7.7	170	460
## 1228	6.2	7.5	7.2	7.5	180	530
## 1229	6.3	7.7	7.3	7.7	210	521
## 1230	4.0	6.1	6.5	6.8	220	438
## 1231	7.5	7.8	7.1	7.5	60	263
## 1232	5.1	7.2	6.2	7.3	150	421
## 1233	5.3	6.9	7.1	7.8	120	360
## 1234	0.7	7.4	6.4	7.5	130	441
## 1235	5.3	7.6	6.9	7.6	140	322
## 1236	6.2	7.4	7.2	7.9	120	355
## 1237	6.5	7.5	7.2	7.5	180	325
## 1238	7.0	7.8	7.2	7.6	200	331
## 1239	6.5	7.6	7.2	7.8	195	290
## 1240	6.1	7.5	7.3	7.6	180	460
## 1241	6.4	7.5	7.2	7.6	190	363
## 1242	6.9	7.7	7.2	7.4	160	228
## 1243	6.8	7.8	7.1	7.6	130	380
## 1244	7.1	7.7	7.1	7.6	140	430
## 1245	6.8	7.5	6.9	7.5	140	374
## 1246	6.6	7.8	4.1	7.4	52	79
## 1247	6.8	7.5	2.8	3.9	137	495
## 1248	6.5	7.7	6.7	7.2	44	636
## 1249	7.1	8.1	7.1	7.6	59	85

## 1250	7.0	7.9	7.1	7.7	62	89
## 1251	6.4	8.1	6.7	7.2	51	69
## 1252	6.9	8.0	6.7	7.1	38	58
## 1253	7.2	8.2	6.9	7.5	43	55
## 1254	6.8	8.1	6.7	7.1	44	65
## 1255	6.9	8.0	6.9	7.3	43	55
## 1256	7.0	7.9	6.7	7.7	35	40
## 1257	7.2	8.2	7.5	7.9	81	130
## 1258	3.0	9.0	6.5	7.3	38	500
## 1259	7.2	8.4	7.1	7.4	47	72
## 1260	2.3	5.5	6.7	7.1	160	385
## 1261	5.0	6.8	7.0	7.4	160	318
## 1262	2.0	5.8	7.1	7.5	240	415
## 1263	2.4	5.6	6.9	7.5	255	518
## 1264	6.8	8.8	6.7	7.2	34	58
## 1265	2.0	4.2	6.7	7.6	165	368
## 1266	4.8	6.8	6.9	7.4	122	284
## 1267	6.7	8.0	6.9	7.4	35	55
## 1268	6.9	7.8	2.3	3.0	650	1430
## 1269	6.8	8.0	6.5	8.1	168	368
## 1270	7.1	8.7	5.0	5.7	55	118
## 1271	7.0	7.8	7.1	7.8	77	97
## 1272	7.1	7.8	6.8	7.6	73	101
## 1273	6.9	8.0	6.9	7.5	58	88
## 1274	7.0	7.9	6.9	7.6	67	78
## 1275	6.8	8.0	6.9	7.4	20	85
## 1276	6.5	6.5	8.1	8.1	20200	20200
## 1277	6.6	6.6	8.0	8.0	29800	29800
## 1278	2.0	2.0	7.4	7.4	2080	2080
## 1279	6.4	6.4	6.9	6.9	186	186
## 1280	4.2	7.3	7.2	8.0	488	680
## 1281	2.0	4.0	7.0	7.8	652	1013
## 1282	2.2	5.3	7.0	7.8	730	989
## 1283	1.9	4.0	7.2	7.8	690	1129
## 1284	1.1	4.9	7.0	7.6	663	1398
## 1285	1.4	5.1	7.2	7.9	597	1093
## 1286	1.2	4.1	7.1	7.7	703	1255
## 1287	1.0	3.4	7.1	7.7	687	1258
## 1288	1.6	5.0	7.1	7.7	607	994
## 1289	1.8	4.2	7.1	7.8	702	1080
## 1290	1.0	5.8	6.8	8.0	589	1218
## 1291	1.4	6.2	7.0	7.9	520	1147
## 1292	1.8	4.2	7.3	7.8	670	1078
## 1293	1.0	3.3	7.3	7.9	792	1182
## 1294	0.3	1.4	7.0	7.9	874	1411
## 1295	3.6	6.1	7.9	8.0	424	789
## 1296	2.8	3.8	7.7	8.0	565	1030
## 1297	4.7	8.2	8.0	9.8	866	1144
## 1298	3.9	6.7	8.0	8.6	190	3600
## 1299	5.2	6.2	7.9	8.6	310	730
## 1300	4.5	6.9	8.0	8.7	320	440
## 1301	4.9	7.1	8.4	8.6	400	560
## 1302	6.1	8.0	5.7	7.6	26	380
## 1303	6.1	7.3	5.7	7.2	39	184

## 1304	6.1	7.1	6.0	7.1	49	1088
## 1305	6.0	7.6	6.2	7.2	74	1504
## 1306	6.0	7.5	6.0	7.3	34	140
## 1307	6.6	7.4	5.6	7.5	33	138
## 1308	6.0	7.6	5.9	7.1	129	771
## 1309	5.4	6.7	7.4	8.1	321	1100
## 1310	5.3	7.4	6.2	7.3	132	1081
## 1311	7.0	7.5	5.6	7.4	33	492
## 1312	5.3	7.9	6.3	7.3	77	675
## 1313	6.0	7.0	6.0	7.1	55	263
## 1314	5.8	6.8	6.0	7.0	65	414
## 1315	5.8	7.0	6.8	7.9	62	924
## 1316	6.1	6.7	7.0	7.4	124	169
## 1317	5.4	6.3	6.9	7.6	131	187
## 1318	5.0	6.1	6.8	7.6	118	178
## 1319	5.1	6.3	6.8	7.7	125	187
## 1320	5.8	6.9	6.8	7.5	125	179
## 1321	6.1	6.7	6.7	7.7	115	173
## 1322	5.9	6.9	6.7	7.6	119	188
## 1323	6.0	6.8	6.9	7.4	125	179
## 1324	5.9	7.0	7.0	7.7	117	178
## 1325	6.2	7.3	6.4	7.1	117	163
## 1326	5.6	6.8	6.9	7.3	114	171
## 1327	5.0	6.9	6.7	7.4	119	202
## 1328	6.0	6.7	6.9	7.5	127	174
## 1329	5.7	6.9	7.0	7.6	92	167
## 1330	5.5	6.6	6.7	7.3	112	157
## 1331	5.3	6.8	6.5	7.6	117	167
## 1332	5.8	6.6	6.8	7.6	121	181
## 1333	5.2	6.8	6.5	7.0	112	179
## 1334	5.2	6.7	6.7	7.9	131	177
## 1335	5.3	6.7	6.4	7.7	119	175
## 1336	5.0	6.7	6.3	7.5	112	164
## 1337	4.3	6.3	6.4	7.3	111	182
## 1338	5.2	6.8	6.8	7.8	118	180
## 1339	5.3	6.7	6.3	7.6	119	174
## 1340	5.1	6.5	6.7	7.4	118	169
## 1341	5.0	6.2	6.7	7.8	115	161
## 1342	Min	Max	Min	Max	Min	Max
## 1343	5.9	8.2	7.0	8.2	242	400
## 1344	5.2	8.3	6.7	8.5	240	612
## 1345	6.4	8.4	7.1	8.8	243	400
## 1346	6.0	7.5	7.1	8.4	330	648
## 1347	5.0	8.3	6.9	8.5	221	410
## 1348	5.8	8.8	7.0	8.3	228	420
## 1349	6.6	7.9	7.6	8.5	104	251
## 1350	2.7	6.7	6.7	8.3	19214	25108
## 1351	4.8	6.1	6.7	8.0	15824	17943
## 1352	6.4	7.7	7.5	8.6	131	406
## 1353	6.6	7.5	7.4	8.9	112	277
## 1354	4.8	7.4	7.7	8.5	147	295
## 1355	4.3	5.4	7.0	8.1	192	4174
## 1356	4.1	6.1	6.8	8.2	1989	7700
## 1357	1.2	8.9	7.3	8.5	383	9900

## 1358	6.3	7.3	7.2	8.9	192	391
## 1359	5.0	9.3	6.3	7.0	41	1843
## 1360	5.7	7.5	7.1	8.3	433	55690
## 1361	6.0	7.9	6.2	7.1	48	12430
## 1362	6.1	8.1	6.3	7.3	53	108
## 1363	5.0	8.1	6.2	7.4	11	118
## 1364	4.7	8.2	6.3	7.5	45	172
## 1365	6.0	8.7	6.7	7.7	48	243
## 1366	4.8	8.7	6.4	6.9	37	181
## 1367	5.5	7.5	6.5	7.4	68	38080
## 1368	6.4	8.0	6.9	8.0	16071	57820
## 1369	5.4	7.5	6.7	8.0	5109	57580
## 1370	5.7	7.9	6.5	7.8	4189	54750
## 1371	5.4	7.7	6.3	7.2	67	41300
## 1372	5.9	8.0	7.0	8.0	17240	57360
## 1373	1.6	6.9	6.2	7.2	69	9700
## 1374	0.3	4.6	6.5	7.8	68	263
## 1375	3.4	7.0	6.6	7.5	296	35940
## 1376	4.0	7.5	6.1	7.7	15340	55020
## 1377	4.4	7.5	6.6	7.5	57	124
## 1378	5.4	7.9	6.3	7.9	49	83
## 1379	6.0	7.4	7.0	8.1	5860	57450
## 1380	7.0	8.2	6.2	6.9	47	128
## 1381	4.1	7.3	6.6	7.2	173	31920
## 1382	5.2	6.8	6.9	7.9	791	53380
## 1383	4.7	7.2	6.5	7.8	799	52840
## 1384	4.5	7.6	6.1	7.1	76	14950
## 1385	0.3	6.1	7.2	7.8	678	2093
## 1386	5.2	7.6	6.8	8.0	350	53210
## 1387	0.3	3.7	7.3	7.9	957	1866
## 1388	0.3	5.2	7.2	7.9	694	2003
## 1389	5.0	7.1	7.2	8.4	662	53452
## 1390	0.3	6.2	7.3	8.4	1269	5348
## 1391	1.5	5.9	7.5	8.4	1004	1919
## 1392	7.3	9.0	8.0	8.5	146	540
## 1393	4.8	6.8	7.2	8.0	306	10470
## 1394	6.7	7.2	7.5	8.0	244	340
## 1395	0.3	2.5	7.4	8.2	648	1520
## 1396	5.7	7.2	7.7	8.2	351	51950
## 1397	5.6	6.7	7.4	8.0	376	3102
## 1398	5.3	6.7	7.3	8.0	3131	41450
## 1399	3.6	7.0	7.2	8.2	733	7936
## 1400	5.2	7.2	7.1	8.4	412	3578
## 1401	4.2	6.4	7.4	8.5	570	12886
## 1402	0.3	5.3	7.3	7.9	498	1307
## 1403	4.9	7.4	7.7	8.2	392	568
## 1404	0.3	6.2	7.2	9.1	760	1261
## 1405	5.2	8.1	6.9	8.2	382	575
## 1406	4.2	6.7	7.3	7.8	470	1378
## 1407	5.4	5.9	8.0	8.2	1850	2280
## 1408	5.1	5.9	8.1	8.3	2030	2240
## 1409	5.2	5.8	8.0	8.2	2630	2950
## 1410	4.2	5.6	6.9	7.5	439	905
## 1411	8.1	9.2	6.8	8.4	423	516

## 1412	7.2	9.1	7.2	8.0	218	676
## 1413	7.1	8.8	7.0	8.0	272	626
## 1414	7.2	8.9	7.1	8.1	343	499
## 1415	4.4	8.8	7.3	7.9	300	3935
## 1416	7.1	8.7	7.0	8.1	286	622
## 1417	7.4	7.8	7.1	7.7	61	110
## 1418	7.2	7.8	6.9	7.6	53	117
## 1419	6.9	7.8	6.7	7.6	122	144
## 1420	7.0	7.7	6.8	7.5	126	158
## 1421	7.3	7.9	6.6	7.5	130	172
## 1422	6.6	7.6	6.7	7.4	118	148
## 1423	6.1	7.1	6.1	6.8	158	205
## 1424	5.7	7.4	6.1	7.3	158	201
## 1425	6.2	7.5	6.2	7.3	165	196
## 1426	5.3	7.1	6.7	7.8	149	183
## 1427	6.3	7.9	6.1	7.6	59	100
## 1428	6.5	7.6	6.5	7.8	53	112
## 1429	6.0	6.8	5.4	6.8	89	124
## 1430	6.3	7.7	6.8	7.9	120	454
## 1431	6.4	7.9	6.7	8.1	124	393
## 1432	6.1	8.2	6.4	7.9	38	187
## 1433	6.3	8.2	6.6	7.8	45	37650
## 1434	3.2	8.1	6.4	7.8	40	86
## 1435	7.6	8.2	6.0	8.0	48	106
## 1436	1.2	8.0	5.4	8.3	99	3200
## 1437	6.2	7.5	6.5	7.9	138	37100
## 1438	5.8	6.9	6.4	7.9	9551	9990
## 1439	6.5	7.3	6.3	7.2	180	220
## 1440	1.2	8.2	7.3	7.8	75	123
## 1441	0.5	5.1	6.0	7.6	92	4613
## 1442	1.5	6.8	5.9	7.4	79	152
## 1443	6.0	7.1	6.2	7.9	160	202
## 1444	1.7	8.0	6.8	7.9	367	55200
## 1445	4.7	7.9	7.2	8.4	135	801
## 1446	6.2	7.2	5.4	6.8	60	94
## 1447	0.3	6.5	5.3	7.7	255	420
## 1448	6.9	8.1	6.3	8.2	54	109
## 1449	6.0	8.2	5.5	7.5	47	117
## 1450	5.7	7.5	6.1	7.3	51	96
## 1451	6.2	8.2	5.7	7.1	114	205
## 1452	4.5	7.7	6.3	8.4	644	1116
## 1453	6.0	8.0	6.5	7.6	42	86
## 1454	5.9	7.7	6.2	7.9	133	47200
## 1455	7.5	8.4	6.1	7.4	21	46
## 1456	6.3	8.2	5.9	7.7	36	82
## 1457	6.1	7.6	5.1	6.9	106	121
## 1458	5.2	7.9	6.7	7.7	81	235
## 1459	5.1	7.9	6.9	7.4	120	185
## 1460	6.3	7.7	6.5	7.5	72	158
## 1461	4.3	8.0	6.2	7.8	84	150
## 1462	3.6	8.1	6.1	7.2	38	65
## 1463	4.5	7.8	6.0	7.9	63	54800
## 1464	6.0	8.2	6.1	7.2	51	76
## 1465	4.7	8.2	6.3	7.9	55	3390

## 1466	4.2	7.7	6.4	7.5	82	49200
## 1467	5.9	7.1	5.8	6.9	43	77
## 1468	5.9	7.0	5.6	6.8	63	112
## 1469	6.0	6.9	6.3	7.2	165	201
## 1470	6.2	7.1	6.3	7.5	44	59
## 1471	5.1	6.3	6.4	7.4	62	93
## 1472	6.3	7.5	6.5	7.3	56	83
## 1473	4.8	5.9	6.1	7.1	54	71
## 1474	6.4	7.7	6.9	7.4	43	78
## 1475	6.7	8.2	6.3	8.6	34	47
## 1476	5.7	8.2	6.0	7.9	37	162
## 1477	6.5	8.0	6.2	7.5	39	58
## 1478	5.8	8.2	5.9	7.7	41	72
## 1479	5.9	8.2	6.2	9.4	70	18920
## 1480	5.6	8.2	6.2	7.8	42	14030
## 1481	6.0	8.2	5.9	8.9	37	67
## 1482	3.9	8.0	6.2	7.9	88	11100
## 1483	6.4	8.0	6.5	7.6	37	89
## 1484	5.3	7.8	5.5	7.9	45	80
## 1485	5.8	7.8	5.6	7.8	117	170
## 1486	4.7	7.6	6.2	7.8	239	48700
## 1487	7.3	7.9	6.3	7.9	61	95
## 1488	6.3	7.8	6.5	7.9	56	95
## 1489	2.3	5.4	6.8	7.6	91	11011
## 1490	5.2	8.1	7.3	7.6	64	118
## 1491	4.1	7.9	6.8	7.8	90	50000
## 1492	6.0	7.9	6.6	7.9	225	49800
## 1493	6.1	6.9	5.6	6.9	52	101
## 1494	5.9	7.5	7.0	8.6	71	3072
## 1495	5.8	7.6	7.1	8.0	71	192
## 1496	6.7	7.5	6.7	8.3	80	245
## 1497	6.8	7.4	7.0	8.2	70	194
## 1498	4.2	7.5	7.1	8.0	118	919
## 1499	6.0	6.6	7.6	7.9	503	999
## 1500	4.3	7.5	7.0	8.1	120	9756
## 1501	3.9	7.5	6.8	8.2	65	10850
## 1502	5.9	7.5	6.9	8.9	68	158
## 1503	3.4	7.1	7.3	8.1	72	59470
## 1504	1.0	6.8	6.9	7.8	338	33170
## 1505	6.1	7.4	7.2	8.0	77	8827
## 1506	4.0	7.5	6.8	7.8	54	3326
## 1507	6.4	7.4	6.9	7.9	59	343
## 1508	5.0	7.5	6.9	7.9	54	6578
## 1509	5.0	7.6	6.8	8.1	55	6493
## 1510	6.4	7.5	7.2	8.0	55	500
## 1511	5.1	7.6	6.8	7.8	61	2867
## 1512	6.6	7.6	7.3	8.0	53	1644
## 1513	6.5	7.5	7.1	8.1	110	1221
## 1514	6.3	7.3	6.7	7.8	116	27230
## 1515	5.2	7.3	6.8	8.1	112	30230
## 1516	3.7	7.1	6.4	8.2	140	36250
## 1517	5.3	7.4	6.9	8.2	108	203
## 1518	6.8	7.6	6.8	8.1	109	185
## 1519	6.5	7.3	7.7	8.5	72	246

## 1520	6.8	7.6	6.6	8.1	126	204
## 1521	6.7	7.5	6.8	8.5	96	168
## 1522	6.6	7.5	6.7	8.2	141	189
## 1523	6.8	7.5	7.6	8.5	72	178
## 1524	6.2	7.4	6.6	7.8	45	142
## 1525	6.5	7.3	7.0	7.8	57	111
## 1526	6.3	7.8	6.7	8.1	52	130
## 1527	6.8	7.7	7.3	7.6	190	210
## 1528	7.2	7.3	6.9	7.5	130	320
## 1529	6.1	7.7	7.0	7.5	140	280
## 1530	7.1	7.5	7.2	7.8	115	250
## 1531	5.2	7.4	6.8	7.6	130	330
## 1532	6.8	7.6	7.2	7.6	150	250
## 1533	4.9	7.7	6.5	7.5	190	342
## 1534	6.4	7.5	7.2	7.6	150	310
## 1535	7.4	7.5	7.4	7.6	80	150
## 1536	6.9	7.6	7.3	7.8	140	310
## 1537	3.4	7.4	6.5	7.6	120	830
## 1538	6.4	7.5	7.2	7.5	170	409
## 1539	7.0	7.8	7.3	7.7	160	395
## 1540	5.4	7.4	6.9	7.7	150	460
## 1541	7.0	7.5	7.2	7.4	150	410
## 1542	7.1	7.5	7.2	7.6	130	210
## 1543	6.9	7.5	6.9	7.6	365	480
## 1544	6.7	7.6	6.9	7.4	150	270
## 1545	6.7	7.5	7.2	7.4	190	320
## 1546	7.1	7.7	7.2	7.7	150	250
## 1547	6.3	7.5	7.0	7.4	180	429
## 1548	7.0	7.4	7.0	7.4	120	320
## 1549	3.5	4.9	6.4	7.3	320	880
## 1550	3.2	4.7	6.4	7.3	330	815
## 1551	5.2	6.8	6.6	7.3	140	370
## 1552	5.2	7.4	6.6	7.4	140	450
## 1553	6.9	7.5	7.3	7.7	150	204
## 1554	7.4	8.0	7.4	7.7	80	240
## 1555	6.6	7.7	7.3	7.8	152	260
## 1556	7.2	7.3	7.4	7.5	170	200
## 1557	7.0	7.8	7.3	7.5	200	290
## 1558	6.6	15.0	7.0	7.6	190	369
## 1559	5.4	7.3	5.2	7.6	220	380
## 1560	7.3	7.6	7.5	7.9	140	230
## 1561	6.9	7.5	6.7	7.5	80	280
## 1562	6.6	7.4	7.0	7.4	170	330
## 1563	7.0	7.8	7.1	7.5	68	105
## 1564	7.0	7.8	7.1	7.4	64	92
## 1565	7.1	8.0	7.0	7.4	62	79
## 1566	7.0	7.8	6.7	7.3	50	91
## 1567	7.0	7.6	3.1	3.7	138	421
## 1568	6.8	8.0	5.7	6.4	44	90
## 1569	7.0	8.0	6.7	7.0	38	70
## 1570	6.8	7.5	6.8	7.2	37	68
## 1571	7.0	7.9	6.7	7.1	33	44
## 1572	6.1	7.2	6.7	7.1	50	95
## 1573	7.2	8.2	7.8	8.1	80	121

## 1574	6.8	8.2	6.8	7.1	35	56
## 1575	6.8	8.7	7.0	7.3	40	50
## 1576	6.7	7.9	6.7	7.4	36	52
## 1577	6.5	7.8	6.8	7.3	52	68
## 1578	0.8	4.1	6.8	7.4	210	715
## 1579	0.3	3.7	6.8	7.4	268	503
## 1580	1.0	3.9	6.9	7.5	295	678
## 1581	4.0	6.6	6.8	7.6	192	400
## 1582	0.5	3.8	6.7	7.2	205	432
## 1583	7.0	8.2	6.9	7.5	30	55
## 1584	3.0	6.9	6.7	7.4	135	370
## 1585	6.7	7.6	6.9	7.4	52	83
## 1586	6.9	8.4	7.0	7.4	34	51
## 1587	4.3	8.1	6.7	8.3	98	1113
## 1588	5.1	8.3	6.5	7.9	109	266
## 1589	5.0	7.2	7.1	7.9	116	243
## 1590	6.6	7.5	6.6	7.6	47	190
## 1591	6.5	8.8	6.3	8.6	54	494
## 1592	5.0	11.5	6.1	8.6	117	282
## 1593	4.5	8.6	7.3	8.4	102	378
## 1594	4.7	10.4	6.6	7.8	84	177
## 1595	5.5	7.6	7.2	8.5	132	274
## 1596	4.4	7.6	7.4	9.6	100	266
## 1597	4.5	7.9	7.3	9.4	101	271
## 1598	5.5	7.2	5.7	8.9	22	133
## 1599	5.1	7.7	7.2	8.2	102	291
## 1600	5.2	8.0	7.2	8.1	169	5957
## 1601	6.0	8.4	7.4	8.2	127	365
## 1602	6.0	8.8	7.4	8.4	106	471
## 1603	6.8	7.5	6.8	8.3	74	184
## 1604	6.2	6.8	7.0	8.1	153	497
## 1605	6.7	7.5	7.0	8.3	147	245
## 1606	6.9	7.6	7.4	8.1	123	379
## 1607	6.5	11.0	7.2	8.5	172	792
## 1608	5.0	8.2	7.3	8.3	177	35760
## 1609	6.0	8.5	7.4	8.4	167	12530
## 1610	5.6	11.0	7.4	8.5	185	323
## 1611	5.8	12.0	6.6	8.5	163	44360
## 1612	6.8	7.6	7.3	8.0	133	494
## 1613	6.8	7.4	7.4	8.3	137	416
## 1614	5.1	6.1	7.3	8.0	524	878
## 1615	5.5	5.5	7.9	7.9	23500	23500
## 1616	6.8	6.8	8.0	8.0	24900	24900
## 1617	6.1	6.1	7.9	7.9	20800	20800
## 1618	6.3	6.3	6.9	6.9	8370	8370
## 1619	1.1	3.4	7.3	7.9	597	1252
## 1620	1.3	3.8	7.3	7.9	538	995
## 1621	2.0	4.0	7.0	7.8	491	995
## 1622	2.6	3.4	7.1	7.7	564	910
## 1623	2.3	3.8	7.2	7.8	529	998
## 1624	4.0	6.0	7.3	7.9	426	883
## 1625	1.2	3.4	7.3	7.9	598	1090
## 1626	0.9	5.2	6.9	7.9	477	828
## 1627	4.2	6.8	7.5	7.9	390	728

## 1628	1.2	3.5	7.1	7.9	568	1128
## 1629	3.7	9.9	7.3	8.1	490	1207
## 1630	2.1	4.6	7.1	7.6	824	1722
## 1631	2.9	4.2	7.2	7.8	461	792
## 1632	2.6	4.2	7.1	7.9	452	855
## 1633	4.0	6.5	7.2	7.9	361	713
## 1634	2.1	3.8	7.0	7.9	502	932
## 1635	1.4	4.0	7.0	7.8	552	1155
## 1636	5.2	7.0	7.0	8.4	653	653
## 1637	6.3	8.2	6.3	8.5	157	601
## 1638	5.4	8.6	6.5	8.0	26	98
## 1639	5.9	8.2	6.4	8.0	49	274
## 1640	5.5	7.6	6.5	8.1	175	1507
## 1641	6.0	9.0	6.3	7.7	37	258
## 1642	6.7	8.1	6.6	7.9	105	974
## 1643	5.3	8.3	6.5	7.7	37	123
## 1644	6.0	8.4	6.4	7.7	33	120
## 1645	5.7	8.5	6.5	8.0	94	436
## 1646	5.0	9.0	7.1	8.3	126	387
## 1647	6.0	8.0	6.1	8.0	83	220
## 1648	5.6	7.6	6.5	7.8	85	270
## 1649	5.7	8.0	6.9	7.9	105	321
## 1650	5.1	6.0	6.7	7.3	135	185
## 1651	5.1	6.3	6.8	7.6	129	199
## 1652	5.0	6.0	7.0	7.5	137	194
## 1653	5.2	5.9	6.9	7.5	129	283
## 1654	5.1	6.7	6.9	7.5	141	183
## 1655	5.1	6.0	6.8	7.4	135	199
## 1656	5.1	6.6	6.6	7.3	127	198
## 1657	5.1	6.9	7.3	7.7	131	189
## 1658	5.1	6.9	7.1	7.6	150	263
## 1659	5.0	5.7	6.8	7.5	141	257
## 1660	5.0	5.9	6.9	7.9	138	248
## 1661	5.1	6.5	6.3	7.3	148	194
## 1662	5.1	6.8	6.5	7.3	124	181
## 1663	5.0	6.5	7.0	7.4	133	196
## 1664	5.1	6.3	6.9	7.4	139	194
## 1665	5.0	5.9	6.8	7.4	134	189
##	BODMin	BODMax	NitrateMin	NitrateMax	FecalColiformMin	FecalColiformMax
## 1	1.1	2.4	1.26	3.23	7	29
## 2	1.3	2.3	0.45	2.15	11	27
## 3	1.0	5.0	0.20	1.92	10	150
## 4	1.6	24.0	0.12	2.06	49	1600
## 5	1.0	20.0	0.66	2.90	130	710
## 6	2.0	22.0	0.15	5.40	17	540
## 7	1.0	8.0	0.40	4.20	40	920
## 8	1.0	5.0	0.29	2.40	32	1600
## 9	1.0	3.0	0.12	1.60	50	920
## 10	1.0	1.0	0.30	1.81	17	1600
## 11	1.0	12.0	0.05	1.60	21	250
## 12	0.4	5.1	0.02	0.52	200	17000
## 13	0.5	2.3	0.02	0.23	45	7900
## 14	0.6	3.2	0.05	0.74	450	4900
## 15	0.4	2.3	0.07	2.61	230	7900

## 16	0.4	3.0	0.01	0.23	450	4900
## 17	0.2	4.3	0.01	0.40	2300	24000
## 18	0.5	2.8	0.02	0.33	130	3500
## 19	0.5	2.1	0.02	0.29	78	3500
## 20	0.4	6.8	BDL	0.82	23	1300
## 21	0.1	3.3	0.06	0.52	780	2400
## 22	0.1	2.0	0.02	0.26	330	2400
## 23	0.5	3.9	0.08	1.35	780	17000
## 24	0.6	2.3	0.01	0.22	230	7900
## 25	0.9	5.1	0.01	8.93	790	4900
## 26	0.9	2.8	0.05	0.44	2200	220000
## 27	0.4	3.5	0.03	0.27	450	7900
## 28	0.6	4.8	0.38	3.72	4900	24000
## 29	1.3	3.3	0.05	0.80	200	7900
## 30	1.1	4.2	0.08	0.67	130	1300
## 31	0.5	16.8	0.10	3.06	11000	54000
## 32	1.3	8.4	0.11	43.76	130	1700
## 33	1.2	3.6	0.04	0.88	230	13000
## 34	1.5	3.6	0.10	1.44	130	7900
## 35	1.4	2.5	0.05	0.09	2	58
## 36	1.9	3.1	0.04	0.67	17	130
## 37	0.3	1.3	0.05	0.56	2	34
## 38	1.6	3.4	0.16	0.89	2	130
## 39	4.0	34.0	0.11	0.88	110	1400
## 40	0.9	2.6	0.11	0.34	2	41
## 41	3.2	148.0	0.20	0.40	11	22
## 42	1.2	3.1	0.04	0.13	94	542
## 43	1.2	3.0	0.02	3.92	84	542
## 44	1.1	2.5	0.01	0.52	48	542
## 45	1.5	2.0	0.04	0.10	2	542
## 46	1.2	16.0	0.28	1.05	12	46
## 47	0.2	31.0	0.17	0.82	49	130
## 48	1.9	67.0	0.21	1.46	11	22
## 49	1.1	8.0	0.02	0.24	4	8
## 50	2.5	21.0	0.02	0.86	40	400
## 51	0.3	1.3	0.09	1.08	2	12
## 52	16.0	40.0	<NA>	<NA>	<NA>	<NA>
## 53	20.0	40.2	<NA>	<NA>	<NA>	<NA>
## 54	16.0	290.0	<NA>	<NA>	7000	25300
## 55	3.0	6.0	<NA>	<NA>	210	2100
## 56	20.0	48.0	<NA>	<NA>	<NA>	<NA>
## 57	0.2	0.4	0.29	1.17	8	21
## 58	0.1	0.3	0.22	0.91	11	17
## 59	0.3	3.2	0.22	0.91	11	21
## 60	0.1	0.3	0.30	0.96	12	17
## 61	10.0	160.0	0.14	3.12	12	350
## 62	0.1	0.6	0.20	0.87	10	17
## 63	2.3	6.5	0.04	0.80	35	550
## 64	2.0	4.0	0.40	1.48	8	94
## 65	1.0	4.0	0.11	2.30	8	94
## 66	2.3	3.0	0.02	0.80	25	550
## 67	1.0	3.0	0.09	1.88	17	920
## 68	1.0	3.0	0.44	1.80	17	920
## 69	0.5	4.0	0.08	1.94	600	6300

## 70	0.8	1.8	0.08	0.55	400	5800
## 71	0.0	2.3	0.01	0.53	14	400
## 72	0.5	1.2	0.04	1.30	30	700
## 73	1.2	104.0	0.10	0.76	110	260
## 74	0.3	1.6	0.07	0.57	500	1800
## 75	0.5	3.0	0.02	1.94	110	700000
## 76	3.7	52.0	0.50	3.00	1420	19500
## 77	0.2	1.3	0.01	2.14	90	370
## 78	1.0	2.1	0.10	1.16	520	1300000
## 79	0.2	2.4	0.03	0.53	800	2000
## 80	0.6	2.2	0.01	0.27	10	700
## 81	0.1	3.3	0.01	1.02	110	580
## 82	1.1	2.0	0.18	1.10	110	210
## 83	0.4	2.5	0.03	0.27	90	200
## 84	0.0	2.0	0.05	0.05	220	800
## 85	1.0	1.4	0.12	0.56	100	180
## 86	0.9	1.0	0.05	0.05	80	240
## 87	1.0	1.6	0.52	0.89	110	190
## 88	0.2	1.8	0.16	0.86	110	490
## 89	0.3	3.0	0.02	168.00	16	255
## 90	0.0	5.0	0.02	0.50	12	260
## 91	0.0	1.0	0.05	0.50	8	480
## 92	0.4	3.0	0.06	1.10	24	940
## 93	0.5	2.2	0.01	0.21	14	188
## 94	0.1	3.0	0.00	0.80	19	184
## 95	0.2	2.5	0.01	0.30	14	212
## 96	0.8	4.6	0.21	5.37	140	5800
## 97	0.5	0.8	0.05	0.05	80	270
## 98	1.1	1.8	0.05	0.05	5	540
## 99	0.6	0.9	0.05	0.05	100	690
## 100	1.2	2.8	0.90	1.40	110	200
## 101	1.0	2.0	0.20	0.96	110	200
## 102	0.3	13.0	0.04	0.50	31	290
## 103	0.6	3.8	0.01	0.52	19	300
## 104	1.1	6.4	0.02	0.54	60	260
## 105	0.6	3.6	0.02	2.00	8	120
## 106	0.3	3.6	0.05	1.40	12	1100
## 107	0.7	4.5	0.07	1.30	40	6000
## 108	0.0	1.4	0.02	0.58	90	4200
## 109	0.2	2.7	0.06	0.87	200	760
## 110	0.7	2.9	0.05	0.72	100	800
## 111	0.1	2.0	0.05	0.80	48	820
## 112	0.2	3.0	0.05	1.05	10	720
## 113	0.1	3.9	0.04	0.84	72	760
## 114	0.3	3.9	0.21	0.74	3	960
## 115	0.3	2.2	0.01	0.30	20	350
## 116	0.1	2.4	0.01	0.54	16	216
## 117	0.0	3.1	0.01	0.50	24	448
## 118	0.3	3.0	0.01	0.20	22	352
## 119	0.2	3.2	0.00	0.50	15	936
## 120	0.5	2.8	0.01	0.50	12	272
## 121	0.3	2.7	0.04	1.56	16	780
## 122	0.4	3.0	0.21	0.52	4	100
## 123	1.1	3.0	0.10	0.73	10	286

## 124	0.3	6.6	0.21	0.99	40	580
## 125	0.5	2.6	0.08	0.66	400	6300
## 126	0.7	5.1	0.20	2.10	240	7000
## 127	0.6	3.7	0.01	3.37	640	6300
## 128	0.4	2.5	0.02	1.50	80	70000
## 129	1.6	6.4	0.05	1.40	480	6300
## 130	1.2	5.2	0.10	1.40	280	4300
## 131	0.2	2.2	0.13	0.60	14	124
## 132	0.6	2.5	0.13	0.90	36	400
## 133	0.4	3.0	0.18	0.65	6	280
## 134	0.0	1.1	0.03	0.41	2	400
## 135	1.0	1.4	0.22	0.78	100	180
## 136	1.0	1.4	0.28	0.62	80	290
## 137	0.2	2.0	0.04	0.48	100	9200
## 138	0.2	0.9	0.05	0.05	40	82
## 139	0.2	2.9	0.18	2.62	8	920
## 140	3.2	75.0	0.50	7.30	22	170
## 141	3.0	3.4	0.30	2.10	2	110
## 142	3.0	4.0	0.10	1.10	5	46
## 143	3.0	4.0	0.20	1.40	8	32
## 144	3.0	4.0	0.10	12.00	13	94
## 145	3.0	3.6	0.20	1.50	2	27
## 146	3.0	4.0	0.30	1.20	17	46
## 147	3.0	4.0	0.20	1.10	2	79
## 148	2.0	3.4	0.12	1.58	6	33
## 149	6.0	250.0	0.30	4.30	540	1600
## 150	3.2	10.0	0.10	2.70	14	130
## 151	2.0	3.4	0.11	1.08	8	110
## 152	3.0	4.0	0.10	0.70	7	94
## 153	3.0	6.0	0.20	2.60	7	79
## 154	3.0	4.0	0.20	1.10	8	70
## 155	2.8	5.0	<NA>	<NA>	2	11
## 156	5.0	18.0	0.60	2.74	23	110
## 157	2.8	3.6	0.10	0.40	17	79
## 158	2.0	3.0	0.18	1.05	5	26
## 159	3.0	15.0	0.10	7.10	13	94
## 160	3.0	5.0	0.20	1.60	8	26
## 161	3.0	4.0	0.30	1.40	9	27
## 162	3.0	4.0	0.30	1.30	17	110
## 163	3.0	4.0	0.20	1.30	2	274
## 164	3.0	4.0	0.20	1.10	4	33
## 165	2.0	2.8	0.12	1.34	11	79
## 166	2.0	3.2	0.18	1.22	17	120
## 167	2.0	3.4	0.14	1.62	13	94
## 168	2.2	3.2	0.15	1.16	8	120
## 169	3.0	4.0	0.10	4.30	2	22
## 170	3.0	4.0	0.10	4.30	8	27
## 171	3.0	4.0	0.10	4.30	8	33
## 172	3.0	3.4	0.20	1.20	5	23
## 173	3.0	3.2	0.10	1.20	8	22
## 174	2.0	3.2	0.13	1.42	8	49
## 175	2.0	2.4	0.10	1.04	2	21
## 176	3.7	3.9	<NA>	<NA>	<NA>	<NA>
## 177	2.9	12.1	<NA>	<NA>	<NA>	<NA>

## 178	2.2	2.6	<NA>	<NA>	<NA>	<NA>
## 179	2.6	2.9	<NA>	<NA>	<NA>	<NA>
## 180	13.1	16.2	<NA>	<NA>	<NA>	<NA>
## 181	13.6	120.0	<NA>	<NA>	<NA>	<NA>
## 182	2.1	28.0	<NA>	<NA>	<NA>	<NA>
## 183	6.0	6.4	<NA>	<NA>	<NA>	<NA>
## 184	2.2	2.3	<NA>	<NA>	<NA>	<NA>
## 185	2.6	3.7	<NA>	<NA>	<NA>	<NA>
## 186	1.3	2.7	<NA>	<NA>	<NA>	<NA>
## 187	1.6	3.6	<NA>	<NA>	<NA>	<NA>
## 188	1.0	1.6	<NA>	<NA>	<NA>	<NA>
## 189	1.6	2.0	<NA>	<NA>	<NA>	<NA>
## 190	1.0	1.2	<NA>	<NA>	<NA>	<NA>
## 191	3.6	5.6	<NA>	<NA>	<NA>	<NA>
## 192	1.5	3.6	<NA>	<NA>	<NA>	<NA>
## 193	5.4	9.5	<NA>	<NA>	<NA>	<NA>
## 194	2.4	4.9	<NA>	<NA>	<NA>	<NA>
## 195	1.2	1.4	<NA>	<NA>	<NA>	<NA>
## 196	1.6	127.0	<NA>	<NA>	<NA>	<NA>
## 197	1.2	1.5	<NA>	<NA>	<NA>	<NA>
## 198	1.4	2.9	<NA>	<NA>	<NA>	<NA>
## 199	1.1	1.7	<NA>	<NA>	<NA>	<NA>
## 200	1.0	1.0	<NA>	<NA>	<NA>	<NA>
## 201	0.9	1.0	<NA>	<NA>	<NA>	<NA>
## 202	1.6	2.2	<NA>	<NA>	<NA>	<NA>
## 203	0.9	1.0	<NA>	<NA>	<NA>	<NA>
## 204	1.2	1.4	<NA>	<NA>	<NA>	<NA>
## 205	1.0	3.6	<NA>	<NA>	<NA>	<NA>
## 206	1.5	1.7	<NA>	<NA>	<NA>	<NA>
## 207	1.3	2.3	<NA>	<NA>	<NA>	<NA>
## 208	1.3	2.1	<NA>	<NA>	<NA>	<NA>
## 209	2.4	8.0	0.29	0.88	70	350
## 210	3.2	5.8	0.15	0.26	4	2100
## 211	2.0	5.2	0.23	0.89	27	2200
## 212	1.0	2.3	0.35	1.22	39	320
## 213	1.2	2.3	0.38	0.51	27	380
## 214	1.6	2.6	0.46	1.68	8	32
## 215	1.2	2.5	0.22	0.68	6	35
## 216	1.0	2.5	0.30	0.54	4	36
## 217	1.8	2.6	0.36	1.50	10	46
## 218	1.4	2.7	0.39	0.96	8	40
## 219	1.1	2.2	0.24	0.67	4	25
## 220	1.0	1.7	0.40	3.60	9	39
## 221	1.1	2.5	0.22	1.40	8	35
## 222	1.0	1.7	0.25	0.63	6	38
## 223	10.0	45.0	0.56	4.30	7000	33000
## 224	9.0	28.0	0.91	5.03	3300	18000
## 225	7.5	85.0	0.46	4.88	22000	48000
## 226	18.5	80.0	0.45	4.79	14000	74000
## 227	1.4	2.5	0.25	0.51	6	40
## 228	18.0	95.0	0.33	1.97	14000	39000
## 229	8.8	30.0	0.37	2.37	6000	25000
## 230	1.1	6.2	0.20	0.62	22	79
## 231	3.2	6.0	0.15	0.40	2	15

## 232	1.7	2.2	0.31	0.51	13	190
## 233	1.3	2.2	0.18	0.57	2	90
## 234	1.4	2.2	0.29	0.52	10	60
## 235	1.4	2.4	0.15	54.00	8	45
## 236	BDL	BDL	0.09	0.09	<NA>	<NA>
## 237	BDL	BDL	0.08	0.08	<NA>	<NA>
## 238	BDL	BDL	0.15	0.15	<NA>	<NA>
## 239	BDL	BDL	0.45	0.45	<NA>	<NA>
## 240	3.0	8.0	0.05	1.38	1700	5600
## 241	7.0	90.0	0.09	1.60	2000	12000
## 242	5.6	17.0	0.07	1.30	2600	7000
## 243	6.1	56.0	0.05	1.50	2100	4700
## 244	4.0	10.0	0.27	1.10	1400	3400
## 245	4.0	19.0	0.07	1.21	1300	3200
## 246	5.8	23.0	0.05	1.60	1400	3400
## 247	7.0	72.0	0.05	3.60	2700	6100
## 248	5.0	28.0	0.09	2.80	1400	5800
## 249	6.0	48.0	0.18	1.40	1700	9400
## 250	5.0	51.0	0.11	1.20	1500	4900
## 251	4.0	48.0	0.50	1.12	1300	3700
## 252	140.0	1299.0	2.80	18.00	20000	340000
## 253	1.1	1.8	0.98	2.28	4	11
## 254	1.0	1.7	0.98	2.62	3	11
## 255	1.1	1.7	0.98	2.28	3	20
## 256	1.0	3.2	0.02	0.42	17	90
## 257	1.0	3.4	0.03	0.26	11	70
## 258	1.0	3.3	0.02	0.29	22	70
## 259	1.4	2.9	0.01	0.14	17	70
## 260	1.0	2.9	0.02	0.22	11	40
## 261	1.0	2.8	0.01	0.16	4	50
## 262	2.0	3.2	0.14	0.98	9	700
## 263	1.0	3.6	0.07	0.38	26	70
## 264	1.0	3.2	0.02	0.19	2	50
## 265	1.9	3.1	0.05	0.19	30	70
## 266	1.9	4.0	0.02	0.31	14	70
## 267	1.0	3.5	0.03	0.34	4	90
## 268	0.1	0.9	0.23	0.67	170	220
## 269	1.0	2.4	0.18	0.66	170	220
## 270	0.4	1.9	0.10	0.70	170	220
## 271	0.7	2.4	0.23	0.77	170	220
## 272	0.3	1.3	0.95	2.83	170	220
## 273	0.8	1.9	0.13	0.69	170	220
## 274	0.7	1.9	0.63	1.73	150	210
## 275	0.1	1.4	0.10	0.64	170	220
## 276	0.2	1.7	0.13	0.62	170	220
## 277	0.5	1.9	0.16	0.63	170	220
## 278	0.5	1.7	0.15	0.53	170	220
## 279	0.9	2.3	0.41	1.04	170	220
## 280	0.3	1.3	0.23	1.08	140	220
## 281	0.5	1.5	0.27	1.00	170	220
## 282	0.5	1.4	0.13	1.33	140	220
## 283	0.7	2.0	0.15	0.52	170	220
## 284	0.4	1.8	0.57	1.57	170	220
## 285	0.3	1.6	0.29	1.16	170	220

## 286	0.5	1.4	0.20	0.78	170	220
## 287	0.4	1.6	0.11	0.66	140	220
## 288	0.2	1.4	0.12	0.64	170	220
## 289	0.4	1.4	0.13	0.80	170	220
## 290	0.4	1.9	0.11	0.60	170	220
## 291	0.6	2.1	0.16	0.60	170	220
## 292	1.1	2.9	0.34	1.43	170	210
## 293	0.1	1.1	0.15	0.83	170	220
## 294	1.2	1.8	1.26	17.11	7	75
## 295	0.9	2.2	1.27	14.53	9	45
## 296	1.0	4.0	0.10	1.32	2	130
## 297	2.0	28.0	0.16	2.20	26	350
## 298	1.6	38.0	0.18	1.84	26	350
## 299	1.8	108.0	0.15	1.80	33	280
## 300	1.0	6.0	0.10	1.80	34	920
## 301	1.0	6.0	0.15	1.42	29	350
## 302	1.0	5.0	0.12	1.66	34	350
## 303	1.0	4.3	0.10	1.90	27	1200
## 304	1.0	24.0	0.17	2.30	46	1600
## 305	0.9	4.7	0.02	1.64	780	14000
## 306	0.7	3.5	BDL	1.35	450	13000
## 307	BDL	4.6	BDL	0.55	200	2300
## 308	0.5	4.5	BDL	0.45	13	4900
## 309	0.4	2.0	BDL	1.01	450	13000
## 310	0.2	2.1	BDL	0.28	1100	13000
## 311	0.4	2.2	BDL	4.00	200	4900
## 312	BDL	2.4	BDL	0.65	78	4900
## 313	0.5	3.7	BDL	0.04	20	2400
## 314	0.5	2.3	BDL	0.49	130	24000
## 315	BDL	4.2	BDL	0.56	130	4900
## 316	0.6	2.8	0.04	0.73	1300	13000
## 317	BDL	3.5	BDL	0.73	780	4900
## 318	0.5	4.8	0.06	3.23	200	4900
## 319	0.4	6.7	BDL	1.24	3300	24000
## 320	BDL	2.2	BDL	1.46	780	13000
## 321	BDL	5.6	0.15	0.63	3300	24000
## 322	BDL	3.8	BDL	1.38	450	4900
## 323	BDL	5.8	BDL	1.42	130	4900
## 324	0.6	5.8	0.74	20.61	11000	35000
## 325	1.0	5.0	0.22	3.02	450	13000
## 326	BDL	4.2	BDL	2.29	450	4900
## 327	BDL	3.5	BDL	13.13	200	4900
## 328	0.7	4.4	BDL	1.31	450	24000
## 329	0.9	2.0	0.03	0.18	2	22
## 330	1.4	4.3	0.15	0.80	210	1600
## 331	0.4	1.3	0.05	0.44	17	70
## 332	0.9	3.3	0.31	4.53	94	920
## 333	3.1	42.0	0.16	1.48	220	1100
## 334	1.1	3.1	0.14	0.33	22	63
## 335	7.6	221.0	0.20	0.48	2	1600
## 336	1.2	9.0	0.09	0.16	3	120
## 337	1.1	4.0	0.05	3.92	3	94
## 338	0.8	2.0	0.02	0.22	3	84
## 339	0.7	1.9	0.03	0.20	3	212

## 340	1.2	21.3	0.38	1.16	130	1600
## 341	3.0	42.0	0.09	1.58	94	220
## 342	0.3	7.7	0.03	1.32	2	70
## 343	0.2	4.5	0.04	0.09	2	12
## 344	3.7	42.1	0.42	1.46	400	1600
## 345	0.3	0.5	0.42	0.84	4	14
## 346	1.9	2.5	0.05	0.35	7	12
## 347	0.4	6.8	2.27	2.55	11	280
## 348	4.2	28.0	<NA>	<NA>	<NA>	<NA>
## 349	20.0	36.0	<NA>	<NA>	<NA>	<NA>
## 350	14.0	110.0	<NA>	<NA>	1200	7400
## 351	5.0	16.0	<NA>	<NA>	700	1300
## 352	16.0	32.0	<NA>	<NA>	<NA>	<NA>
## 353	2.6	3.0	BDL	0.92	80	350
## 354	2.4	2.9	BDL	0.45	170	900
## 355	2.0	3.0	0.09	1.70	17	220
## 356	1.0	2.0	0.09	1.42	14	170
## 357	1.3	5.2	0.39	1.26	200	5400
## 358	0.5	1.8	0.12	1.00	310	2600
## 359	0.5	1.9	0.04	0.50	49	2200
## 360	0.3	1.8	0.02	1.20	48	4600
## 361	1.2	1.8	0.44	0.54	49	140
## 362	0.3	0.7	0.03	0.50	1000	2000
## 363	1.0	2.5	0.27	1.30	79	2600
## 364	6.1	58.0	1.20	2.90	1700	25000
## 365	0.2	1.4	0.14	0.56	120	640
## 366	0.5	2.1	0.27	0.88	1200	5400
## 367	0.4	0.7	0.02	0.42	600	2400
## 368	0.2	3.0	0.00	0.20	0	700
## 369	0.4	1.8	0.00	0.89	110	540
## 370	1.2	1.8	0.09	0.96	60	140
## 371	0.2	2.5	0.00	0.20	0	400
## 372	0.9	4.7	0.00	0.50	480	3200
## 373	1.1	1.8	0.46	0.62	70	180
## 374	0.7	1.3	0.00	0.50	200	310
## 375	1.1	1.8	0.76	0.88	91	170
## 376	0.5	1.4	0.00	0.93	120	400
## 377	0.4	3.6	0.20	2.00	6	120
## 378	0.5	1.0	0.00	0.60	10	790
## 379	0.5	1.0	0.00	0.80	40	2700
## 380	0.2	2.2	0.01	0.80	100	1200
## 381	0.5	3.2	0.00	0.58	220	580
## 382	0.5	2.7	0.00	0.56	150	400
## 383	0.5	3.0	0.00	0.40	150	430
## 384	2.0	6.7	0.50	8.60	20	4300
## 385	0.6	5.4	<NA>	<NA>	310	500
## 386	0.8	3.9	<NA>	<NA>	490	920
## 387	0.8	1.5	<NA>	<NA>	100	310
## 388	1.2	1.8	0.72	1.40	17	180
## 389	1.4	1.8	0.48	0.92	60	160
## 390	0.4	3.6	0.01	2.30	0	490
## 391	0.8	2.7	0.06	1.50	40	630
## 392	1.2	3.2	0.14	1.20	10	310
## 393	1.2	3.2	0.00	8.00	10	150

## 394	0.2	3.6	0.00	5.00	6	120
## 395	0.7	3.5	0.11	4.90	2100	11000
## 396	0.5	2.3	0.00	2.00	100	1800
## 397	0.2	2.5	0.00	1.62	20	2240
## 398	0.1	2.3	0.00	0.54	48	2000
## 399	0.5	2.5	0.00	1.30	40	1100
## 400	0.5	2.7	0.00	1.21	30	1400
## 401	0.3	2.7	0.00	0.33	140	790
## 402	0.1	3.9	0.00	0.95	80	1500
## 403	0.1	2.9	0.00	0.39	220	490
## 404	0.5	2.9	0.00	0.59	170	450
## 405	0.5	2.3	0.00	0.26	250	540
## 406	0.6	2.8	0.00	0.58	200	370
## 407	0.8	3.0	0.00	0.56	220	450
## 408	0.5	2.4	0.00	7.30	270	350
## 409	0.2	2.0	0.00	0.64	39	1700
## 410	0.6	5.1	0.03	4.10	0	360
## 411	0.4	2.9	0.04	1.32	0	480
## 412	0.4	3.6	0.02	1.75	0	700
## 413	0.5	2.4	0.22	1.96	400	3900
## 414	1.5	5.7	0.19	1.86	200	4000
## 415	2.0	3.4	0.19	2.00	540	3400
## 416	0.8	4.8	0.00	1.08	40	2700
## 417	2.7	6.7	0.12	2.00	790	4600
## 418	2.0	6.5	0.10	2.00	200	4300
## 419	0.3	1.2	0.02	0.43	2	33
## 420	0.2	4.4	0.13	0.23	4	100
## 421	0.6	3.5	0.05	8.20	0	640
## 422	0.5	0.9	0.00	1.80	110	1200
## 423	1.1	2.0	0.52	0.86	40	150
## 424	1.1	2.2	0.32	0.56	70	310
## 425	0.2	2.0	0.03	0.90	100	1700
## 426	0.9	2.1	<NA>	<NA>	70	150
## 427	0.5	2.3	0.00	0.42	10	790
## 428	0.2	2.0	0.00	1.51	12	11600
## 429	1.0	4.3	0.00	1.42	22	65000
## 430	3.0	3.0	0.90	0.90	21	21
## 431	3.5	3.5	0.90	0.90	26	26
## 432	1.2	5.6	2.00	35.00	<NA>	<NA>
## 433	0.8	4.3	26.00	35.00	<NA>	<NA>
## 434	0.3	4.3	25.00	40.00	<NA>	<NA>
## 435	2.0	4.1	30.00	45.00	<NA>	<NA>
## 436	3.7	68.7	30.00	40.00	<NA>	<NA>
## 437	0.4	74.9	26.00	35.00	<NA>	<NA>
## 438	0.7	8.2	27.00	40.00	<NA>	<NA>
## 439	0.6	4.3	30.00	40.00	<NA>	<NA>
## 440	3.6	3.6	38.00	38.00	<NA>	<NA>
## 441	2.8	2.8	40.00	40.00	<NA>	<NA>
## 442	3.4	3.4	35.00	35.00	<NA>	<NA>
## 443	1.2	4.1	27.00	45.00	<NA>	<NA>
## 444	0.9	4.9	30.00	46.00	<NA>	<NA>
## 445	1.2	5.3	26.00	55.00	<NA>	<NA>
## 446	2.1	5.4	28.50	50.00	<NA>	<NA>
## 447	0.8	78.1	28.00	48.00	<NA>	<NA>

## 448	0.8	6.9	25.00	45.00	<NA>	<NA>
## 449	0.1	81.0	30.00	50.00	<NA>	<NA>
## 450	0.6	4.5	27.00	45.00	<NA>	<NA>
## 451	1.0	5.8	26.00	40.00	<NA>	<NA>
## 452	0.6	3.3	28.00	40.00	<NA>	<NA>
## 453	0.6	3.8	30.00	40.00	<NA>	<NA>
## 454	0.1	5.6	30.00	35.00	<NA>	<NA>
## 455	3.1	3.1	<NA>	<NA>	<NA>	<NA>
## 456	0.6	6.2	25.00	40.00	<NA>	<NA>
## 457	1.7	5.7	30.00	45.00	2	2
## 458	0.5	5.9	25.00	50.00	<NA>	<NA>
## 459	0.8	6.8	30.00	50.00	<NA>	<NA>
## 460	5.4	5.4	<NA>	<NA>	<NA>	<NA>
## 461	0.2	5.2	25.00	40.00	<NA>	<NA>
## 462	0.7	5.5	25.00	50.00	<NA>	<NA>
## 463	4.0	5.7	0.35	1.00	39	2100
## 464	3.2	4.3	0.11	0.52	4	12
## 465	1.9	4.8	0.24	0.50	920	2200
## 466	1.0	2.0	0.33	0.53	110	940
## 467	1.0	2.0	0.31	0.48	64	210
## 468	1.4	2.3	0.27	1.75	14	27
## 469	1.2	1.8	0.21	0.60	11	26
## 470	1.2	1.8	0.46	1.02	12	21
## 471	1.8	2.6	0.10	1.35	17	27
## 472	1.2	2.1	0.28	1.15	17	26
## 473	1.4	1.9	0.32	0.94	11	23
## 474	1.0	1.8	0.37	0.51	16	22
## 475	1.3	2.5	0.28	1.03	15	23
## 476	1.0	1.5	0.27	0.97	12	22
## 477	18.0	35.0	0.30	1.75	3100	70000
## 478	9.5	27.0	1.35	7.30	2600	79000
## 479	25.0	46.0	0.29	0.83	35000	120000
## 480	1.5	40.0	0.24	0.74	29000	140000
## 481	1.4	2.4	0.13	0.92	17	36
## 482	24.0	40.0	0.15	1.22	22000	84000
## 483	8.9	22.0	0.54	1.73	12000	49000
## 484	1.0	4.3	0.30	0.56	28	79
## 485	3.5	5.2	0.08	0.40	0	4
## 486	1.3	2.0	0.07	0.57	27	41
## 487	1.2	1.8	0.16	0.32	9	22
## 488	1.2	2.0	0.44	0.54	21	27
## 489	1.2	2.0	0.30	0.51	17	27
## 490	0.6	2.6	0.05	7.75	2	3500
## 491	0.9	3.1	0.12	1.62	2	1700
## 492	0.8	2.8	0.34	1.31	170	3500
## 493	0.5	2.1	0.25	2.79	45	1700
## 494	0.4	1.5	0.24	6.61	2	2400
## 495	0.6	1.6	0.15	0.83	200	35000
## 496	1.0	2.8	0.18	1.83	490	35000
## 497	0.3	1.4	0.13	1.42	20	2400
## 498	0.5	1.5	0.10	4.31	20	700
## 499	0.2	1.6	0.11	1.18	20	2400
## 500	1.0	2.0	0.01	1.33	<NA>	<NA>
## 501	<NA>	<NA>	0.06	0.06	<NA>	<NA>

## 502	2.0	2.0	0.06	0.06	<NA>	<NA>
## 503	20.0	20.0	0.07	0.07	<NA>	<NA>
## 504	<NA>	<NA>	0.41	0.41	<NA>	<NA>
## 505	6.0	18.0	0.20	1.10	2100	4100
## 506	14.0	110.0	0.52	1.80	3300	13000
## 507	11.0	31.0	0.28	1.80	3100	6300
## 508	14.0	58.0	0.21	1.40	2700	21000
## 509	8.0	35.0	0.25	1.90	1700	5800
## 510	8.0	18.0	0.29	1.30	1400	2700
## 511	11.0	22.0	0.30	1.40	1700	4100
## 512	10.0	50.0	0.49	2.20	2400	6300
## 513	8.0	39.0	0.28	1.70	1700	4900
## 514	12.0	68.0	0.42	2.10	2600	11000
## 515	9.0	44.0	0.33	1.30	1400	8400
## 516	7.0	40.0	0.31	1.30	1300	5800
## 517	95.0	260.0	1.54	12.00	230000	1700000
## 518	1.1	2.8	1.70	3.07	4	9
## 519	2.4	2.4	1.76	1.76	28	28
## 520	1.8	1.8	2.06	2.06	20	20
## 521	0.6	2.5	1.84	3.04	4	14
## 522	1.1	1.5	2.46	2.80	7	14
## 523	1.1	1.7	1.84	3.07	4	20
## 524	2.0	88.2	2.62	3.15	20	75
## 525	1.3	5.8	0.08	0.32	4	32
## 526	1.0	6.8	0.07	2.94	6	70
## 527	1.1	6.0	0.05	2.44	8	60
## 528	1.0	6.0	0.07	0.97	14	70
## 529	1.0	6.2	0.02	0.49	6	40
## 530	1.2	5.8	0.01	0.38	4	32
## 531	1.6	8.7	0.12	0.80	17	90
## 532	4.4	4.4	1.85	1.85	2	2
## 533	1.6	5.2	0.07	0.23	16	70
## 534	1.5	6.4	0.01	0.44	4	40
## 535	1.0	6.0	0.09	0.92	14	60
## 536	1.0	5.5	0.15	3.96	7	70
## 537	1.3	7.3	0.21	4.41	12	90
## 538	0.1	0.7	0.15	0.46	110	170
## 539	0.2	1.8	0.14	0.75	110	170
## 540	0.2	1.3	0.16	0.89	110	150
## 541	0.2	0.9	0.22	0.60	130	350
## 542	0.2	0.6	0.31	1.12	110	170
## 543	0.2	0.9	0.17	0.40	110	170
## 544	0.3	0.9	0.28	0.80	120	170
## 545	0.2	0.9	0.15	0.45	120	220
## 546	0.2	0.9	0.20	0.53	110	170
## 547	0.3	0.9	0.16	0.51	110	170
## 548	0.3	1.1	0.16	0.63	120	170
## 549	0.3	1.2	0.30	0.82	110	170
## 550	0.3	1.0	0.16	0.55	110	220
## 551	0.2	1.1	0.20	0.77	110	170
## 552	0.2	0.8	0.17	0.53	110	150
## 553	0.3	0.9	0.15	0.50	110	170
## 554	0.2	0.7	0.29	0.70	110	170
## 555	0.2	1.1	0.22	0.44	110	170

## 556	0.2	1.0	0.15	0.51	110	170
## 557	0.3	0.8	0.20	0.52	110	150
## 558	0.2	0.9	0.15	0.39	110	170
## 559	0.2	0.7	0.13	0.47	110	220
## 560	0.3	1.1	0.15	0.40	130	280
## 561	0.3	1.1	0.15	0.40	94	170
## 562	0.2	1.3	0.28	0.72	110	170
## 563	0.3	1.1	0.16	0.36	110	150
## 564	1.3	2.4	0.70	3.09	4	20
## 565	1.2	2.5	0.23	9.35	7	21
## 566	0.8	2.7	0.24	2.13	4	11
## 567	1.2	2.6	0.96	8.64	15	20
## 568	1.4	2.6	0.16	2.29	7	21
## 569	1.2	2.4	0.38	1.77	7	21
## 570	0.4	1.2	0.09	0.42	BDL	2
## 571	1.2	2.4	0.33	1.07	4	15
## 572	2.2	3.4	0.56	1.40	3	4
## 573	1.4	2.3	0.18	3.26	9	23
## 574	1.7	14.0	0.15	3.30	3	9
## 575	1.3	2.7	0.50	1.60	2	2000
## 576	1.1	2.5	0.60	2.10	300	2100
## 577	2.2	2.8	1.10	1.50	2	2
## 578	1.2	5.2	0.40	3.00	2	730
## 579	16.0	34.0	1.80	3.10	2800	15000
## 580	1.1	2.8	0.60	2.00	2	1500
## 581	1.0	2.9	0.50	2.20	2	360
## 582	1.4	2.8	1.00	4.20	300	910
## 583	1.3	2.3	0.40	2.10	300	2000
## 584	1.5	4.9	0.40	2.50	2	730
## 585	1.3	2.7	0.90	2.60	2	1500
## 586	1.2	2.8	1.00	2.60	2	910
## 587	1.3	2.8	0.60	1.80	2	1100
## 588	18.4	78.0	1.80	4.00	300	2800
## 589	1.2	2.7	0.40	1.80	2	730
## 590	0.4	2.4	0.80	2.10	2	1500
## 591	1.3	3.2	0.20	2.30	300	1100
## 592	1.8	2.8	0.70	1.30	300	1400
## 593	2.0	2.9	0.80	1.80	2	730
## 594	1.8	4.4	1.30	2.40	300	1500
## 595	1.3	3.0	1.30	1.40	2	2
## 596	1.0	2.1	0.70	2.00	720	1500
## 597	1.5	5.0	0.40	2.30	2	1100
## 598	1.1	2.9	0.60	2.00	2	1100
## 599	1.1	2.8	0.30	1.90	300	730
## 600	1.2	2.2	0.50	2.20	2	1500
## 601	1.4	62.0	0.80	3.50	1500	3900
## 602	14.0	34.0	1.90	3.30	2700	24000
## 603	0.4	2.3	0.80	1.90	2	360
## 604	1.4	3.2	0.50	1.30	2	910
## 605	1.1	2.4	1.10	2.40	300	910
## 606	1.0	2.6	0.90	2.40	2	1100
## 607	1.7	5.6	0.11	0.31	1700	28000
## 608	1.7	2.2	0.03	0.45	1100	14000
## 609	1.8	3.1	0.08	0.17	820	35000

## 610	1.8	3.2	0.06	0.27	830	17000
## 611	19.0	19.0	1.45	1.45	2000	2000
## 612	206.0	206.0	1.81	1.81	20000	20000
## 613	94.0	94.0	1.14	1.14	4000	4000
## 614	21.0	21.0	2.36	2.36	2000	2000
## 615	1.0	1.0	0.02	0.02	350000	350000
## 616	BDL	1.0	0.10	0.40	2	9
## 617	BDL	3.0	0.40	2.50	900	1600
## 618	1.5	3.1	0.70	10.00	23	110
## 619	BDL	2.6	0.20	2.50	21	90
## 620	BDL	2.2	0.10	0.50	34	220
## 621	BDL	1.1	0.10	0.40	21	60
## 622	BDL	BDL	0.10	0.60	8	34
## 623	BDL	3.0	0.10	0.30	11	40
## 624	BDL	2.9	0.10	1.60	240	1600
## 625	BDL	2.1	BDL	0.27	110	1300
## 626	BDL	2.9	0.11	0.91	2300	13000
## 627	BDL	BDL	BDL	0.55	2	2400
## 628	BDL	2.9	0.04	0.23	490	4900
## 629	BDL	4.0	0.03	0.75	450	3300
## 630	BDL	4.2	BDL	0.13	68	1700
## 631	BDL	2.8	BDL	0.17	78	2200
## 632	BDL	2.3	0.01	0.16	170	2300
## 633	BDL	2.7	BDL	0.48	490	2300
## 634	BDL	1.8	BDL	1.41	23	1300
## 635	BDL	2.8	0.02	0.31	5	2300
## 636	BDL	1.1	BDL	1.42	20	790
## 637	BDL	5.0	0.01	0.75	2300	11000
## 638	BDL	3.5	BDL	1.11	490	4900
## 639	BDL	2.8	0.01	0.42	78	3300
## 640	BDL	6.0	0.39	45.00	7900	350000
## 641	BDL	3.2	BDL	1.31	130	2300
## 642	BDL	BDL	BDL	1.27	BDL	790
## 643	BDL	2.7	BDL	1.22	45	2100
## 644	BDL	1.7	0.04	0.82	130	2300
## 645	BDL	2.7	BDL	0.12	130	1700
## 646	BDL	2.1	BDL	0.27	790	3500
## 647	BDL	2.1	0.02	1.24	790	4900
## 648	BDL	3.8	0.03	0.35	460	1700
## 649	BDL	6.4	0.04	1.78	130	2300
## 650	BDL	5.0	0.08	0.45	130	2300
## 651	0.9	1.4	0.04	0.25	6	11
## 652	2.0	49.0	0.19	0.89	220	1400
## 653	2.0	43.0	0.13	0.88	70	140
## 654	0.9	2.9	0.02	0.23	6	33
## 655	2.2	258.6	0.22	0.61	17	49
## 656	0.2	1.8	0.10	0.33	6	170
## 657	0.8	5.3	0.45	0.64	26	1600
## 658	2.7	33.0	0.34	2.16	210	1600
## 659	0.7	0.7	0.03	0.09	23	23
## 660	1.0	1.1	0.02	0.28	4	17
## 661	1.6	2.9	0.19	0.39	94	1600
## 662	1.0	28.0	0.03	0.23	6	27
## 663	0.6	24.0	0.02	0.42	17	58

## 664	0.7	1.2	0.02	0.19	6	17
## 665	0.8	0.8	0.95	0.95	4	4
## 666	0.9	2.4	0.04	0.24	BDL	4
## 667	1.8	38.0	0.42	1.06	96	1800
## 668	0.3	0.7	0.36	0.68	2	12
## 669	2.0	90.0	0.06	34.00	2200	250000
## 670	16.0	60.0	-	-	12500	180000
## 671	7.0	60.0	0.11	1.12	22000	340000
## 672	4.0	7.5	0.05	2.96	14000	940000
## 673	16.0	70.0	22.00	22.00	9600	198000
## 674	0.1	0.1	-	-	21	26
## 675	0.1	0.1	-	-	11	27
## 676	0.1	1.8	0.11	0.45	4	26
## 677	0.1	0.2	0.07	1.00	2	20
## 678	0.1	0.3	0.13	1.12	2	37
## 679	0.3	1.0	0.10	1.16	12	22
## 680	0.4	1.7	0.02	0.89	11	33
## 681	0.6	1.8	0.01	0.75	17	43
## 682	0.4	1.6	0.05	1.41	13	26
## 683	0.6	16.0	0.07	2.12	39	1700
## 684	0.5	2.1	0.05	1.32	14	27
## 685	0.1	0.3	0.13	0.72	2	4
## 686	0.1	0.3	0.18	0.97	2	6
## 687	0.1	0.2	-	-	13	40
## 688	-	-	-	-	-	-
## 689	-	-	0.01	0.01	-	-
## 690	0.2	0.2	-	-	-	-
## 691	0.6	3.0	-	-	2	140
## 692	3.4	4.0	-	-	200	230
## 693	3.6	4.0	-	-	200	230
## 694	2.1	2.5	-	-	-	-
## 695	3.0	3.6	-	-	-	-
## 696	2.8	8.0	-	-	2	2
## 697	0.4	0.9	-	-	2	2
## 698	2.6	3.0	-	-	140	200
## 699	2.0	2.8	0.10	0.52	170	350
## 700	2.0	2.6	0.13	0.55	110	300
## 701	2.6	26.0	0.02	0.21	35	350
## 702	2.3	2.6	0.02	0.18	17	900
## 703	2.1	2.3	0.10	0.29	25	250
## 704	2.0	2.0	1.00	1.10	14	500
## 705	2.0	6.0	0.10	1.47	17	110
## 706	2.0	2.0	0.88	1.12	22	170
## 707	2.0	2.0	1.04	1.12	34	140
## 708	1.6	2.4	0.60	0.82	70	280
## 709	1.6	2.6	0.56	0.84	100	320
## 710	1.6	2.8	0.62	0.86	130	480
## 711	1.6	2.4	0.54	0.76	130	410
## 712	0.2	2.5	0.01	0.30	20	150
## 713	0.1	2.1	0.01	0.32	20	170
## 714	1.0	5.5	0.01	0.10	100	310
## 715	0.8	2.9	0.03	2.00	2	260
## 716	1.1	2.6	0.10	7.80	2	1200
## 717	0.3	2.6	0.01	1.63	79	2400

## 718	0.1	1.7	0.01	0.88	100	5600
## 719	0.1	1.5	0.08	0.90	60	4600
## 720	0.1	1.9	0.01	0.83	150	4900
## 721	0.3	2.8	0.01	0.30	17	280
## 722	0.4	2.8	0.16	7.38	170	6300
## 723	0.1	2.4	0.03	2.65	200	4600
## 724	1.2	2.1	0.62	0.98	70	210
## 725	1.4	2.1	1.00	1.60	91	280
## 726	0.8	3.4	0.01	0.80	4	70
## 727	0.6	2.1	0.02	0.85	4	40
## 728	1.2	3.6	0.10	1.80	200	100000
## 729	0.4	4.4	0.03	0.83	120	6300
## 730	1.2	2.1	0.48	0.66	82	280
## 731	0.2	4.6	0.04	1.69	200	12800
## 732	1.2	3.0	0.19	12.04	10	1000
## 733	0.1	2.0	0.12	0.16	2	56
## 734	0.3	1.8	0.01	0.50	16	220
## 735	2.9	10.2	1.00	2.60	3100	190000
## 736	1.1	4.3	0.01	0.10	79	500
## 737	0.8	2.9	0.01	0.32	60	480
## 738	0.3	2.1	0.01	0.25	10	180
## 739	0.9	2.8	0.01	0.34	100	480
## 740	0.7	3.0	0.29	5.84	2	400
## 741	0.3	2.8	0.01	0.30	10	120
## 742	0.2	2.9	0.01	0.31	10	130
## 743	0.1	1.5	0.01	0.60	50	1400
## 744	0.1	2.4	0.01	0.49	180	3400
## 745	0.5	4.6	0.01	0.06	100	910
## 746	0.4	3.0	0.01	0.31	50	610
## 747	0.3	2.4	0.05	0.24	100	790
## 748	0.2	2.8	0.10	0.21	2	84
## 749	1.0	2.9	0.01	0.60	4	220
## 750	0.2	0.9	0.06	0.80	310	2800
## 751	0.5	2.1	0.01	0.21	31	430
## 752	0.2	0.8	0.11	0.48	1100	2700
## 753	0.4	2.9	0.03	0.60	32	390
## 754	0.6	2.7	0.01	0.21	26	240
## 755	0.1	3.9	0.01	0.22	100	600
## 756	0.4	6.5	0.01	0.30	100	480
## 757	1.2	2.2	0.52	0.76	100	280
## 758	0.8	2.5	0.15	0.54	110	380
## 759	0.5	3.9	0.26	0.41	140	460
## 760	0.2	1.7	0.22	1.14	210	42500
## 761	0.8	3.9	0.27	2.25	520	32500
## 762	0.3	1.5	0.14	0.59	170	630
## 763	0.3	2.8	0.07	1.58	170	12000
## 764	0.5	2.6	0.01	1.19	170	3800
## 765	0.4	2.1	0.01	2.63	580	7000
## 766	0.1	1.6	0.01	0.59	240	3500
## 767	0.2	2.9	0.01	2.62	400	4700
## 768	0.3	2.4	0.01	0.52	350	4600
## 769	0.1	2.8	0.05	1.12	40	2700
## 770	0.3	2.7	0.01	0.97	10	3500
## 771	0.2	2.6	0.05	0.50	20	210

## 772	1.0	3.8	0.01	0.57	26	380
## 773	1.0	2.6	0.01	1.42	55	400
## 774	0.3	2.4	0.01	0.23	10	150
## 775	0.1	2.8	0.01	0.32	19	80
## 776	0.3	2.1	0.01	0.35	10	150
## 777	0.8	3.2	0.24	1.90	8	48
## 778	0.1	3.3	0.01	0.66	32	490
## 779	0.7	2.7	0.01	0.51	5	170
## 780	0.4	6.3	0.01	0.30	100	940
## 781	0.8	1.5	0.39	0.93	2	2
## 782	0.8	1.4	0.52	1.64	2	2
## 783	1.8	2.0	0.61	0.90	2	6
## 784	1.6	6.0	0.34	1.60	2	140
## 785	1.5	2.3	0.13	1.52	2	18
## 786	0.5	1.2	0.01	0.44	2	2
## 787	0.5	1.4	0.69	1.03	2	2
## 788	1.8	2.4	0.14	1.99	2	32
## 789	0.7	1.4	0.14	0.96	2	18
## 790	3.0	4.9	0.10	3.05	2	170
## 791	3.0	10.0	0.10	3.23	2	210
## 792	3.0	4.0	0.10	3.30	11	220
## 793	3.0	4.0	0.10	1.50	2	220
## 794	3.0	7.0	0.10	1.30	2	540
## 795	3.0	6.5	2.30	14.10	2	350
## 796	3.0	4.0	0.10	0.80	2	350
## 797	3.0	9.0	0.30	2.80	6	540
## 798	3.0	15.0	0.30	1.50	2	79
## 799	3.0	5.4	0.10	8.70	2	350
## 800	3.0	50.0	0.10	2.40	13	3500
## 801	2.0	12.0	0.02	0.91	2	920
## 802	3.0	4.0	0.20	1.70	2	220
## 803	3.0	5.8	0.20	1.80	9	350
## 804	3.0	11.0	0.40	2.70	17	34
## 805	3.0	4.0	0.20	6.82	8	220
## 806	3.0	4.0	0.20	1.70	9	240
## 807	3.0	4.0	0.20	1.70	2	540
## 808	3.0	6.6	0.20	1.70	14	110
## 809	3.0	64.0	0.20	22.10	2	170
## 810	1.8	13.0	0.16	0.70	6	1600
## 811	2.0	11.0	0.13	1.58	4	110
## 812	2.0	12.0	0.05	1.00	2	540
## 813	2.0	50.0	0.22	1.40	5	920
## 814	2.0	12.0	0.22	1.50	7	920
## 815	3.0	8.0	BDL	1.00	2	280
## 816	3.0	4.0	BDL	0.90	2	170
## 817	3.0	11.0	BDL	1.60	2	280
## 818	3.0	4.0	0.10	1.20	4	350
## 819	3.0	4.0	0.30	1.70	2	110
## 820	3.0	4.0	0.10	2.10	11	79
## 821	3.0	4.0	0.10	1.80	4	110
## 822	3.0	4.0	0.20	1.40	23	240
## 823	1.8	2.6	0.06	0.64	2	8
## 824	1.8	4.0	0.05	1.08	2	12
## 825	1.8	2.8	0.03	0.94	2	9

## 826	1.4	6.9	0.02	0.03	2	75
## 827	2.4	4.3	0.02	0.03	2	45
## 828	2.7	4.3	0.03	0.03	2	35
## 829	2.0	4.0	0.03	0.03	2	30
## 830	2.0	4.5	0.02	0.04	2	45
## 831	2.1	4.9	0.03	0.03	2	135
## 832	2.0	3.2	0.03	0.03	2	65
## 833	2.5	2.5	0.03	0.03	-	-
## 834	2.7	3.9	0.02	0.17	2	25
## 835	2.7	3.8	0.03	0.03	2	50
## 836	2.0	3.8	0.03	0.03	2	20
## 837	2.3	3.8	0.03	0.03	2	45
## 838	2.7	4.4	0.02	0.03	2	55
## 839	1.8	3.8	0.03	0.03	2	10
## 840	2.6	4.2	0.02	0.03	55	152
## 841	2.8	4.4	<NA>	<NA>	2	30
## 842	2.6	4.7	0.03	0.03	2	40
## 843	2.9	3.9	0.02	0.03	2	45
## 844	2.6	4.1	0.03	0.03	2	55
## 845	2.1	4.0	0.02	0.03	2	35
## 846	5.2	6.0	0.03	0.04	2	180
## 847	5.1	6.2	0.03	0.03	2	205
## 848	2.0	5.6	0.03	0.03	2	90
## 849	2.7	4.0	0.02	0.03	2	5
## 850	3.1	4.8	0.02	0.03	2	85
## 851	1.8	4.5	0.03	0.03	2	65
## 852	2.7	3.8	0.03	0.03	2	60
## 853	2.9	4.4	0.03	0.04	2	60
## 854	2.6	3.7	<NA>	<NA>	2	10
## 855	2.7	4.9	0.03	0.03	2	20
## 856	2.2	3.8	0.02	0.03	2	75
## 857	2.6	4.4	0.02	0.04	2	55
## 858	1.5	2.3	0.29	0.51	8	32
## 859	1.5	7.1	0.25	0.52	12	38
## 860	1.0	1.8	0.40	0.57	170	1400
## 861	1.1	2.0	0.31	0.47	79	1100
## 862	3.2	5.5	0.20	0.30	2	14
## 863	3.3	5.5	0.20	0.36	2	2
## 864	1.2	1.8	0.42	0.85	13	49
## 865	3.2	6.5	0.46	0.65	210	2400
## 866	1.2	1.9	0.28	0.47	11	26
## 867	1.3	1.8	0.32	0.54	110	410
## 868	1.0	1.7	0.45	0.56	17	94
## 869	1.1	1.9	0.04	0.44	2	140
## 870	1.2	2.2	0.37	1.52	14	32
## 871	1.6	2.4	0.33	0.84	11	31
## 872	1.5	1.8	0.50	1.10	12	26
## 873	1.0	1.5	0.35	0.69	2	27
## 874	1.6	7.4	1.00	1.70	17	36
## 875	1.1	1.9	0.35	0.47	12	32
## 876	1.4	2.2	0.54	1.00	6	24
## 877	21.0	50.0	0.36	6.24	21000	110000
## 878	10.5	40.0	0.64	6.45	8400	110000
## 879	24.0	56.0	0.28	1.68	23000	210000

## 880	26.0	40.0	0.27	0.74	27000	230000
## 881	1.2	2.7	0.23	0.96	10	27
## 882	1.0	1.7	0.28	0.56	2	26
## 883	1.2	2.8	0.40	0.64	11	34
## 884	24.0	44.0	0.33	1.00	22000	130000
## 885	13.0	31.0	0.45	2.27	9100	79000
## 886	3.3	5.8	0.40	1.10	120	700
## 887	1.4	5.4	0.31	0.84	38	120
## 888	BDL	BDL	0.37	0.37	-	-
## 889	BDL	BDL	0.34	0.34	-	-
## 890	0.1	0.1	0.34	0.34	-	-
## 891	BDL	BDL	0.24	0.24	-	-
## 892	4.0	17.0	0.69	4.63	1700	4300
## 893	10.0	46.0	0.70	1.61	2100	7000
## 894	5.0	50.0	0.78	1.94	2100	9200
## 895	10.0	42.0	0.60	0.98	2100	7000
## 896	11.0	28.0	0.72	1.13	2200	7000
## 897	4.0	21.0	0.58	0.91	1400	6300
## 898	10.0	26.0	0.74	1.30	1700	6800
## 899	9.0	25.0	0.72	1.32	1700	5800
## 900	8.0	21.0	0.62	0.93	1400	3400
## 901	12.0	35.0	0.69	1.29	2300	5800
## 902	8.0	64.0	0.51	1.02	1700	8400
## 903	6.0	60.0	0.32	0.93	1500	4100
## 904	11.0	59.0	0.75	1.21	1700	6300
## 905	14.0	66.0	0.66	1.49	3100	9200
## 906	30.0	210.0	0.98	2.51	49000	210000
## 907	-	-	1.54	2.10	7	93
## 908	-	-	1.54	2.36	7	75
## 909	-	-	1.54	2.20	7	75
## 910	2.0	2.0	0.54	1.05	BDL	18
## 911	BDL	3.3	0.01	1.06	2	18
## 912	BDL	5.2	0.01	1.58	2	41
## 913	BDL	5.5	0.06	1.64	4	63
## 914	BDL	6.4	0.02	2.44	4	50
## 915	BDL	4.2	0.02	1.16	2	17
## 916	BDL	3.5	0.05	0.98	2	17
## 917	BDL	5.5	0.06	1.08	2	70
## 918	2.4	5.8	0.08	1.40	20	84
## 919	BDL	4.0	0.02	1.01	2	13
## 920	2.3	6.4	0.01	9.13	12	70
## 921	BDL	3.0	0.02	1.66	7	23
## 922	BDL	6.0	0.02	2.36	5	70
## 923	2.2	2.2	0.02	0.02	84	84
## 924	1.1	2.7	0.26	2.73	11	170
## 925	1.2	2.5	0.21	1.61	58	120
## 926	0.7	2.5	0.17	2.46	46	110
## 927	0.5	2.6	0.20	1.35	31	130
## 928	1.1	2.6	0.21	1.10	46	120
## 929	0.9	2.8	0.25	1.34	43	150
## 930	0.8	2.0	0.30	111.28	63	120
## 931	0.6	2.8	0.28	1.09	94	150
## 932	0.6	2.7	0.17	1.10	70	150
## 933	0.6	1.9	0.25	2.63	63	170

## 934	0.4	1.9	0.26	2.41	46	150
## 935	0.4	1.9	0.28	1.72	31	140
## 936	0.6	2.9	0.35	1.67	63	170
## 937	0.5	2.5	0.23	0.83	79	150
## 938	1.1	3.8	0.31	2.52	150	350
## 939	0.4	2.3	0.29	3.34	70	170
## 940	0.5	2.4	0.27	1.94	58	170
## 941	0.7	2.1	0.28	1.61	40	170
## 942	0.4	2.1	0.25	2.26	58	170
## 943	0.3	2.5	0.27	2.07	70	170
## 944	0.3	2.3	0.28	1.98	70	210
## 945	0.7	2.2	0.31	2.00	63	170
## 946	0.7	2.2	0.23	2.30	94	170
## 947	0.9	1.9	0.23	1.84	70	130
## 948	1.8	2.9	0.29	2.54	31	130
## 949	1.4	2.9	0.28	2.85	48	130
## 950	Min	Max	Min	Max	Min	Max
## 951	1.2	2.7	0.15	2.05	4	15
## 952	1.1	3.0	0.10	3.93	7	21
## 953	1.0	2.7	0.25	1.72	7	21
## 954	1.7	4.5	0.25	9.50	9	23
## 955	1.2	2.6	0.07	3.65	7	21
## 956	1.0	2.8	0.34	3.09	7	21
## 957	2.2	2.8	0.66	1.30	3	3
## 958	0.5	1.0	0.06	0.43	2	100
## 959	2.2	2.2	0.48	0.48	69	69
## 960	1.2	2.5	0.40	1.00	300	700
## 961	1.3	2.1	0.40	1.20	300	730
## 962	1.3	2.3	0.50	1.00	300	730
## 963	1.2	2.2	0.40	1.40	300	1100
## 964	1.4	2.1	0.50	1.80	2	360
## 965	0.9	2.0	0.40	1.80	300	360
## 966	0.7	2.2	0.40	1.60	2	360
## 967	1.9	2.3	0.30	0.90	2	300
## 968	1.1	2.2	0.60	2.60	300	600
## 969	1.6	2.3	0.60	1.60	300	1500
## 970	1.3	1.7	0.70	0.80	2	2
## 971	2.2	2.7	0.70	1.40	300	730
## 972	1.4	2.3	0.80	1.80	300	610
## 973	1.2	2.1	0.60	1.40	300	730
## 974	1.3	2.1	0.30	1.20	300	910
## 975	1.2	2.3	0.40	1.70	300	730
## 976	1.6	2.2	0.50	1.10	300	730
## 977	1.2	2.2	0.60	1.40	300	1100
## 978	1.3	2.2	0.40	0.80	300	1400
## 979	1.4	2.2	0.60	1.10	300	1500
## 980	1.1	2.5	0.70	1.00	300	1100
## 981	1.4	2.8	0.60	1.10	300	910
## 982	1.3	2.3	0.05	0.74	780	160000
## 983	1.9	3.8	0.05	0.54	2100	160000
## 984	2.5	4.0	0.19	0.57	4000	160000
## 985	1.1	2.0	0.15	0.33	1100	160000
## 986	1.0	1.0	0.10	0.60	7	17
## 987	1.0	4.0	0.30	3.00	500	1600

## 988	1.9	3.0	1.00	2.50	90	130
## 989	1.0	3.0	0.10	3.00	33	170
## 990	1.0	2.0	0.10	0.50	21	280
## 991	1.0	1.0	0.10	3.60	11	40
## 992	1.0	1.0	0.10	1.80	7	40
## 993	1.0	1.0	0.10	1.80	7	22
## 994	1.0	2.8	0.10	3.40	40	1600
## 995	5.5	6.4	9.89	24.98	-	-
## 996	1.0	2.9	0.11	0.72	1300	4900
## 997	1.0	1.9	0.11	1.09	70	130
## 998	1.0	2.3	0.16	0.82	78	2200
## 999	1.0	1.7	0.11	0.78	680	1300
## 1000	1.0	1.0	0.06	0.62	230	2300
## 1001	1.0	2.0	0.11	0.32	1300	2200
## 1002	1.0	2.3	0.11	0.32	220	1700
## 1003	1.0	2.5	0.11	0.38	110	1400
## 1004	1.0	1.2	0.11	0.32	330	2300
## 1005	1.0	2.2	0.11	0.32	270	1700
## 1006	1.0	2.5	0.11	0.32	110	1700
## 1007	1.0	2.5	0.11	0.66	780	4900
## 1008	1.0	1.7	0.11	0.32	780	2300
## 1009	1.0	1.0	0.11	1.50	1100	3300
## 1010	1.0	5.0	0.11	0.74	1300	7900
## 1011	1.0	1.8	0.11	0.54	1300	4900
## 1012	1.0	2.8	0.11	0.69	3100	7900
## 1013	1.0	2.7	0.11	1.13	130	1700
## 1014	1.0	3.0	0.11	0.91	680	2300
## 1015	1.0	6.0	0.11	2.45	4900	13000
## 1016	1.0	2.3	0.11	4.11	1100	2200
## 1017	1.0	2.0	0.11	1.67	330	23000
## 1018	1.0	2.1	0.10	1.46	450	2700
## 1019	1.0	2.2	0.11	10.62	13	330
## 1020	1.0	1.4	0.09	0.71	22	1300
## 1021	1.0	2.0	0.11	1.05	82	330
## 1022	0.2	1.1	0.03	1.12	2	18
## 1023	1.0	2.9	0.21	0.45	90	1600
## 1024	0.5	1.9	0.09	0.43	9	41
## 1025	0.9	2.8	0.06	0.64	11	1600
## 1026	2.1	24.0	0.13	0.57	140	330
## 1027	0.8	2.2	0.22	0.70	12	80
## 1028	2.9	238.0	0.14	1.10	2	20
## 1029	0.8	8.0	0.06	0.16	2	540
## 1030	0.4	1.8	0.06	1.30	6	350
## 1031	0.5	1.5	0.04	0.14	2	46
## 1032	0.3	1.4	0.04	1.33	2	18
## 1033	0.9	19.0	0.14	1.03	25	1600
## 1034	1.7	18.0	0.09	0.53	94	170
## 1035	1.0	20.6	0.08	1.11	2	25
## 1036	1.7	2.9	0.03	0.80	2	2
## 1037	4.7	26.0	0.20	1.48	31	1600
## 1038	0.5	1.3	0.08	0.54	4	25
## 1039	3.0	3.1	0.83	0.86	-	-
## 1040	16.0	54.0	8.00	26.00	48000	210000
## 1041	18.0	52.0	10.00	34.00	64000	250000

## 1042	4.5	58.0	0.03	1.14	2100	49000
## 1043	2.6	6.8	0.28	0.91	200	34000
## 1044	22.0	64.0	6.00	18.00	11000	170000
## 1045	5.3	83.0	0.53	7.31	2000	1300000
## 1046	1.0	2.9	1.41	2.83	400	6800
## 1047	16.0	16.0	3.24	3.24	830	830
## 1048	5.8	7.9	0.51	2.19	170	450
## 1049	7.1	7.1	0.83	0.83	17000	17000
## 1050	0.3	1.0	0.07	1.01	2	31
## 1051	0.5	1.6	0.07	0.91	11	79
## 1052	0.7	2.0	0.01	0.89	12	84
## 1053	0.4	2.0	0.02	0.89	9	58
## 1054	0.4	2.4	0.06	0.86	10	94
## 1055	0.1	9.4	0.14	1.13	2	2
## 1056	0.1	1.0	0.08	1.06	2	2
## 1057	0.1	1.0	0.08	1.70	2	2
## 1058	0.1	1.0	0.08	0.49	2	2
## 1059	0.3	1.7	0.02	0.90	6	22
## 1060	0.3	2.0	0.03	0.85	9	32
## 1061	0.1	1.6	-	-	3	13
## 1062	0.1	1.0	0.02	0.84	2	58
## 1063	0.1	1.0	0.25	0.80	13	40
## 1064	0.2	1.0	0.26	2.06	9	34
## 1065	0.1	1.0	0.56	4.95	17	49
## 1066	0.1	1.0	0.31	0.62	13	38
## 1067	0.1	1.0	0.07	1.40	2	2
## 1068	0.1	1.0	0.06	1.53	2	2
## 1069	0.1	1.0	0.15	0.28	2	2
## 1070	0.1	2.4	-	-	2	21
## 1071	1.0	5.0	0.21	0.64	45	54000
## 1072	1.0	5.7	1.22	1.71	78	2200
## 1073	2.4	2.6	-	-	-	-
## 1074	2.3	2.9	-	-	-	-
## 1075	2.1	4.0	-	-	-	-
## 1076	1.0	1.6	1.44	1.66	-	-
## 1077	3.6	7.0	1.14	1.36	-	-
## 1078	0.8	1.0	-	-	-	-
## 1079	3.2	30.0	-	-	200	230
## 1080	3.4	29.0	-	-	200	200
## 1081	2.4	2.9	0.03	0.06	63	137
## 1082	2.0	2.0	0.50	1.38	2	1600
## 1083	2.0	2.0	0.62	1.29	2	1600
## 1084	2.2	2.6	0.02	0.04	58	122
## 1085	2.0	2.0	0.61	1.17	2	1600
## 1086	2.0	3.0	0.47	1.38	2	1600
## 1087	2.3	2.6	0.03	0.52	4	21
## 1088	2.1	2.6	0.09	0.32	5	25
## 1089	2.1	2.6	0.09	0.43	9	17
## 1090	2.1	2.5	0.01	0.47	7	26
## 1091	2.1	2.3	0.11	0.25	4	21
## 1092	2.1	2.6	0.12	0.69	2	33
## 1093	2.0	3.0	0.10	2.64	94	1100
## 1094	2.0	3.0	0.10	0.80	170	700
## 1095	0.2	7.6	0.04	1.91	27	3500

## 1096	0.2	1.8	0.01	0.96	130	3100
## 1097	0.2	1.0	0.04	0.70	110	6300
## 1098	0.1	1.0	0.01	0.53	32	7000
## 1099	1.4	2.0	0.68	0.78	79	140
## 1100	0.3	1.0	0.01	0.38	490	2500
## 1101	0.1	1.9	0.01	0.94	14	1800
## 1102	1.9	9.0	0.79	1.90	1800	5800
## 1103	0.3	1.0	0.01	0.61	70	270
## 1104	0.3	3.0	0.04	0.66	220	40000
## 1105	0.2	1.0	0.01	0.49	400	2100
## 1106	1.2	3.2	0.01	0.31	100	910
## 1107	0.5	2.1	0.01	0.68	40	240
## 1108	1.6	2.6	0.68	0.87	94	180
## 1109	0.9	2.6	0.01	0.27	100	790
## 1110	1.2	3.1	0.01	0.01	100	200
## 1111	1.7	2.5	0.66	0.92	79	200
## 1112	0.4	2.6	0.01	0.01	100	200
## 1113	1.3	1.8	0.58	0.70	60	110
## 1114	0.7	2.1	0.01	0.55	70	490
## 1115	1.0	5.8	0.03	0.80	8	70
## 1116	0.1	1.0	0.01	0.40	91	2200
## 1117	0.3	1.9	0.02	0.45	150	10000
## 1118	0.3	1.8	0.01	0.37	40	180
## 1119	1.4	2.5	0.01	6.50	70	130
## 1120	0.4	2.3	0.04	0.75	12	81
## 1121	0.5	2.3	0.03	0.41	56	94
## 1122	0.8	3.5	0.04	6.60	24	4700
## 1123	0.1	1.7	0.01	0.01	100	200
## 1124	0.3	2.9	0.01	0.01	100	310
## 1125	1.0	2.1	0.01	0.01	100	210
## 1126	1.3	2.0	0.96	1.30	100	170
## 1127	1.5	2.7	0.68	0.80	60	180
## 1128	0.9	2.3	0.01	0.42	100	1100
## 1129	1.0	2.6	0.01	0.75	100	200
## 1130	0.7	2.4	0.01	2.30	100	580
## 1131	0.3	2.8	0.30	1.86	12	84
## 1132	0.3	5.4	0.01	0.45	10	79
## 1133	0.4	3.6	0.10	0.98	460	41000
## 1134	0.3	1.0	0.01	0.68	91	3100
## 1135	0.1	1.5	0.01	0.20	31	120
## 1136	0.3	1.9	0.01	0.05	40	170
## 1137	0.2	2.9	0.01	0.04	39	170
## 1138	0.5	1.9	0.01	0.19	40	200
## 1139	0.2	2.8	0.01	0.18	40	150
## 1140	0.4	2.7	0.01	0.16	40	340
## 1141	1.2	2.3	0.01	0.37	49	100
## 1142	1.0	2.5	0.01	0.43	24	94
## 1143	1.0	2.5	0.01	0.72	48	100
## 1144	1.1	2.5	0.01	0.26	40	79
## 1145	1.0	2.7	0.01	0.89	48	100
## 1146	1.0	1.9	0.01	0.41	32	84
## 1147	0.1	1.4	0.01	0.10	20	140
## 1148	1.0	2.8	0.18	11.90	2	810
## 1149	1.0	1.8	0.02	6.30	20	3800

## 1150	1.0	1.9	0.23	2.41	60	900
## 1151	0.1	2.9	0.01	0.52	100	1700
## 1152	0.2	2.5	0.01	3.97	24	70000
## 1153	0.2	2.6	0.06	0.45	310	3800
## 1154	0.3	2.8	0.01	0.58	79	1400
## 1155	1.0	4.0	0.01	3.06	12	5800
## 1156	1.2	3.3	0.01	0.66	340	3500
## 1157	0.3	1.8	0.01	0.11	4	70
## 1158	0.3	1.5	0.01	0.10	2	84
## 1159	1.2	2.9	0.18	3.69	40	1200
## 1160	0.4	1.0	0.02	0.49	240	7000
## 1161	1.3	2.9	0.68	0.80	79	180
## 1162	1.5	3.6	0.68	0.74	91	170
## 1163	0.2	2.5	0.01	0.96	34	1000
## 1164	0.8	1.6	0.01	0.01	100	200
## 1165	0.3	2.4	0.01	0.33	40	120
## 1166	0.4	1.8	0.01	1.10	140	5400
## 1167	0.5	4.1	0.01	1.50	200	18000
## 1168	0.5	1.6	0.01	0.13	2	2
## 1169	0.7	2.0	0.10	0.33	2	20
## 1170	1.5	2.0	0.09	1.69	2	6
## 1171	1.2	2.0	0.10	1.00	2	6
## 1172	1.4	3.8	0.12	1.38	2	350
## 1173	0.8	1.8	0.06	1.30	2	2
## 1174	0.6	1.2	0.51	1.71	2	2
## 1175	0.7	1.2	0.61	3.37	2	2
## 1176	0.9	1.4	0.63	1.95	2	2
## 1177	0.7	1.8	0.19	1.42	2	10
## 1178	0.7	1.8	0.21	1.52	2	11
## 1179	3.0	20.0	0.30	1.90	2	170
## 1180	2.6	4.0	0.30	0.80	2	170
## 1181	2.8	4.2	0.30	0.80	2	40
## 1182	2.4	6.0	0.30	10.40	8	110
## 1183	2.6	4.8	0.30	1.10	2	33
## 1184	2.8	3.4	0.30	0.70	5	130
## 1185	2.4	4.0	0.30	1.50	2	170
## 1186	2.8	4.2	0.30	5.30	2	70
## 1187	1.5	2.0	0.30	1.00	4	13
## 1188	3.0	54.0	0.30	5.30	350	92000
## 1189	2.8	12.0	0.30	4.20	2	21
## 1190	1.6	2.8	0.30	2.90	6	17
## 1191	2.4	4.0	0.30	1.60	2	110
## 1192	3.0	4.0	0.30	0.50	2	17
## 1193	2.8	3.8	0.30	0.50	2	33
## 1194	2.8	3.2	0.90	4.27	11	14
## 1195	3.4	12.0	0.30	2.70	2	240
## 1196	2.6	4.4	0.30	1.80	2	40
## 1197	1.4	2.0	0.30	0.60	4	9
## 1198	2.8	12.0	0.30	10.00	2	110
## 1199	2.6	4.2	0.30	1.30	2	110
## 1200	2.4	4.8	0.30	1.10	2	140
## 1201	2.8	5.0	0.30	1.40	2	540
## 1202	2.8	4.4	0.30	1.60	2	220
## 1203	2.8	4.0	0.30	0.50	2	21

## 1204	1.6	2.0	0.30	0.70	4	14
## 1205	1.6	2.0	0.30	0.78	5	17
## 1206	1.6	2.4	0.30	3.00	6	21
## 1207	1.6	2.8	0.30	1.44	6	17
## 1208	2.8	4.0	0.30	0.90	2	21
## 1209	2.8	4.0	0.30	1.70	2	21
## 1210	2.8	4.2	0.30	1.70	2	23
## 1211	2.8	4.0	0.40	1.80	11	17
## 1212	2.6	4.0	0.30	1.20	8	27
## 1213	1.5	2.0	0.30	1.22	4	17
## 1214	1.5	1.8	0.30	1.08	4	12
## 1215	2.8	3.6	-	-	5	50
## 1216	2.7	3.2	-	-	25	60
## 1217	2.5	3.8	-	-	10	55
## 1218	2.6	4.0	-	-	35	75
## 1219	2.8	6.5	-	-	95	200
## 1220	3.2	6.8	-	-	130	235
## 1221	2.7	3.7	-	-	25	55
## 1222	2.7	3.7	-	-	30	55
## 1223	2.8	3.7	-	-	15	75
## 1224	2.7	3.7	-	-	55	135
## 1225	3.0	3.9	-	-	15	65
## 1226	2.6	3.9	-	-	35	60
## 1227	2.6	3.6	-	-	20	40
## 1228	2.6	3.7	-	-	25	50
## 1229	2.0	3.9	-	-	20	65
## 1230	3.9	5.8	-	-	120	320
## 1231	2.5	3.4	-	-	2	2
## 1232	2.7	4.6	-	-	69	155
## 1233	3.6	4.6	-	-	30	45
## 1234	3.0	5.5	-	-	35	289
## 1235	2.5	3.9	-	-	5	45
## 1236	2.3	3.8	-	-	30	55
## 1237	2.7	3.5	-	-	5	30
## 1238	2.7	3.6	-	-	5	15
## 1239	2.4	3.7	-	-	5	15
## 1240	3.1	3.8	-	-	40	75
## 1241	2.7	3.7	-	-	30	55
## 1242	2.7	3.5	-	-	2	5
## 1243	2.5	3.7	-	-	10	25
## 1244	2.8	3.4	-	-	30	55
## 1245	2.8	3.6	-	-	30	55
## 1246	2.2	3.7	0.28	2.51	27	280
## 1247	2.2	3.5	0.11	0.38	2	2
## 1248	2.3	3.4	0.35	0.61	130	340
## 1249	1.2	1.8	0.38	0.79	130	220
## 1250	1.4	1.8	0.34	0.53	63	140
## 1251	1.5	5.0	0.72	1.60	11	23
## 1252	1.2	1.9	0.21	0.40	6	14
## 1253	1.0	1.8	0.34	0.75	2	12
## 1254	1.3	1.9	0.62	0.92	12	33
## 1255	1.2	2.0	0.31	0.52	12	27
## 1256	1.4	4.0	0.58	1.23	14	27
## 1257	1.0	8.2	0.48	0.64	2	6

## 1258	1.1	30.0	0.31	0.45	6	28000
## 1259	1.0	5.0	0.30	0.39	2	4
## 1260	7.2	35.0	0.59	6.40	3800	25000
## 1261	5.0	17.6	0.80	8.41	3100	12000
## 1262	8.0	38.5	0.70	3.70	5800	38000
## 1263	8.5	32.0	0.46	3.21	4100	31000
## 1264	1.0	1.9	0.28	0.68	2	17
## 1265	13.0	35.0	0.31	0.86	7000	27000
## 1266	6.0	18.0	0.70	5.43	2700	13000
## 1267	1.2	2.5	0.29	2.55	27	230
## 1268	2.2	3.5	0.15	0.35	2	2
## 1269	1.4	1.8	0.31	0.65	2	14
## 1270	1.3	2.0	0.38	0.61	7	24
## 1271	1.2	2.1	0.30	0.56	8	31
## 1272	1.4	2.0	0.31	0.59	2	24
## 1273	1.1	1.9	0.32	0.58	17	27
## 1274	1.3	1.8	0.31	0.56	34	79
## 1275	1.2	2.0	0.26	0.48	2	24
## 1276	1.0	1.0	0.38	0.38	-	-
## 1277	1.0	1.0	2.50	2.50	-	-
## 1278	1.0	1.0	1.29	1.29	-	-
## 1279	1.0	1.0	-	-	-	-
## 1280	5.0	18.0	1.02	3.14	1400	4000
## 1281	12.0	27.0	0.80	1.50	2100	4900
## 1282	17.0	44.0	0.80	2.11	2600	5800
## 1283	13.0	37.0	1.00	1.70	2500	5800
## 1284	8.0	53.0	0.70	2.00	1700	6300
## 1285	5.0	34.0	0.72	1.90	1700	4300
## 1286	7.0	43.0	0.90	2.40	2100	5800
## 1287	10.0	32.0	0.90	3.20	2000	7000
## 1288	8.0	28.0	0.50	1.60	1700	4300
## 1289	11.0	28.0	0.60	1.80	1700	6300
## 1290	9.0	42.0	0.60	2.10	2000	4700
## 1291	7.0	37.0	0.50	1.90	1700	3800
## 1292	12.0	35.0	0.80	1.50	2200	4700
## 1293	18.0	42.0	0.70	1.90	2700	5800
## 1294	40.0	111.0	1.70	4.10	94000	210000
## 1295	3.0	42.0	0.56	2.56	2200	38000
## 1296	9.0	21.0	0.23	2.17	3800	130000
## 1297	5.0	14.0	1.24	1.95	2	1100
## 1298	0.7	2.9	0.84	2.37	20	120
## 1299	1.0	1.6	1.02	2.24	93	93
## 1300	1.1	2.8	0.85	1.84	23	75
## 1301	1.4	3.1	0.58	2.29	39	75
## 1302	1.0	13.0	0.02	0.40	2	12
## 1303	1.0	24.0	0.02	0.56	4	22
## 1304	2.0	3.0	0.02	1.79	11	110
## 1305	1.0	3.3	0.02	0.52	17	150
## 1306	1.0	3.0	0.02	0.46	2	17
## 1307	1.0	3.0	0.02	0.86	2	17
## 1308	2.4	5.1	0.02	0.85	41	220
## 1309	2.0	3.0	0.32	1.11	2	45
## 1310	2.2	5.1	0.04	0.91	34	140
## 1311	1.0	3.0	0.02	0.41	2	22

## 1312	1.0	4.0	0.02	0.63	25	110
## 1313	2.2	3.0	0.02	1.82	14	94
## 1314	2.4	3.0	0.02	0.51	21	120
## 1315	2.0	3.1	0.02	0.54	20	110
## 1316	1.0	1.7	0.51	1.16	26	70
## 1317	1.0	2.2	0.52	0.99	38	70
## 1318	1.0	1.7	0.35	1.53	17	49
## 1319	1.2	2.6	1.15	3.25	58	170
## 1320	1.0	1.3	0.79	1.78	21	94
## 1321	0.6	1.4	0.52	1.91	17	58
## 1322	1.0	1.5	0.41	1.19	15	63
## 1323	1.0	1.5	0.90	2.85	22	63
## 1324	0.8	1.3	0.48	1.57	26	48
## 1325	0.6	1.4	0.51	1.27	21	63
## 1326	0.5	1.2	0.54	1.79	15	40
## 1327	0.7	1.7	0.97	3.15	21	79
## 1328	0.9	1.5	1.01	3.01	25	94
## 1329	0.3	1.3	0.55	1.61	17	58
## 1330	0.9	1.3	0.66	1.27	17	47
## 1331	0.9	1.7	0.43	2.19	22	47
## 1332	1.0	1.6	0.52	2.51	22	58
## 1333	0.7	1.3	0.46	1.25	15	32
## 1334	1.0	1.8	1.05	2.29	25	63
## 1335	1.0	1.9	1.19	2.79	17	58
## 1336	0.8	1.8	1.02	2.22	21	63
## 1337	0.7	1.8	0.87	1.97	21	58
## 1338	1.0	1.6	0.47	1.84	17	58
## 1339	0.7	1.3	0.62	1.09	25	48
## 1340	0.6	1.8	0.68	1.87	17	58
## 1341	0.5	1.4	0.57	1.78	21	63
## 1342	Min	Max	Min	Max	Min	Max
## 1343	1.4	2.0	0.72	1.66	7	15
## 1344	1.2	2.5	0.92	3.02	11	21
## 1345	1.2	2.0	0.54	3.60	7	15
## 1346	1.4	2.3	0.33	4.02	11	15
## 1347	1.5	2.2	0.42	2.46	9	21
## 1348	1.5	2.2	0.42	3.05	7	15
## 1349	1.0	1.0	0.30	0.30	5	9
## 1350	1.0	7.1	0.50	0.90	21	53
## 1351	2.6	6.1	0.30	0.50	36	110
## 1352	1.0	1.0	0.30	0.30	35	41
## 1353	1.0	1.0	0.30	0.50	12	18
## 1354	1.0	14.3	0.40	4.00	71	214
## 1355	1.0	14.8	0.90	2.10	340	1600
## 1356	2.3	12.2	0.40	1.10	460	1600
## 1357	1.8	9.1	0.32	33.44	920	17000
## 1358	1.0	1.0	0.30	0.50	4	18
## 1359	1.0	1.9	0.32	0.42	2300	7900
## 1360	1.3	2.1	0.32	0.52	230	2200
## 1361	1.0	1.9	0.32	0.33	230	1700
## 1362	1.0	1.5	0.32	0.32	230	2300
## 1363	1.0	1.9	0.32	0.32	230	1300
## 1364	1.0	1.9	0.32	0.32	220	790
## 1365	1.0	1.9	0.32	1.13	490	1700

## 1366	1.0	1.6	0.32	0.35	110	1700
## 1367	1.0	1.9	0.32	0.43	130	1700
## 1368	1.2	2.0	0.32	0.80	8	230
## 1369	1.0	1.9	0.32	1.02	78	460
## 1370	1.2	1.9	0.32	0.98	110	330
## 1371	1.0	2.0	0.32	0.41	330	1400
## 1372	1.0	1.9	0.32	1.46	78	210
## 1373	1.3	3.6	0.32	0.73	130	4900
## 1374	1.3	6.0	0.32	3.82	2300	35000
## 1375	1.0	3.9	0.32	1.73	45	1700
## 1376	1.0	2.5	0.32	1.02	490	2200
## 1377	1.0	3.3	0.32	0.42	330	2700
## 1378	1.0	1.9	0.32	0.32	5	490
## 1379	1.1	1.9	0.32	0.52	170	1100
## 1380	1.0	1.9	0.32	0.33	230	2400
## 1381	1.0	1.9	0.32	0.57	130	3300
## 1382	1.0	2.0	0.32	3.62	2200	3300
## 1383	1.0	2.4	0.32	0.53	490	2200
## 1384	1.0	1.9	0.32	0.36	230	1700
## 1385	1.0	22.0	0.02	1.28	33	1600
## 1386	1.0	2.1	0.05	1.09	10	38
## 1387	1.0	20.0	0.30	0.83	79	1600
## 1388	1.0	23.0	0.30	1.32	70	1600
## 1389	1.3	2.7	0.30	1.01	15	920
## 1390	1.0	22.0	0.32	0.34	11	220
## 1391	1.4	6.0	0.32	0.90	2	47
## 1392	1.0	3.0	0.30	2.21	12	13000
## 1393	1.0	2.2	0.30	0.64	38	1600
## 1394	1.0	1.0	0.30	0.52	8	28
## 1395	1.9	33.0	0.32	1.67	1600	1600
## 1396	1.0	2.4	0.05	1.07	10	27
## 1397	1.0	1.5	0.32	0.43	33	920
## 1398	1.0	1.5	0.33	0.66	10	27
## 1399	1.6	2.9	0.01	0.90	220	920
## 1400	1.0	2.2	0.04	0.90	94	430
## 1401	1.8	2.3	0.30	1.00	2	2
## 1402	3.0	40.0	1.04	3.59	1300	4900
## 1403	2.8	5.4	0.30	0.88	200	9400
## 1404	2.8	42.0	0.10	0.30	800	41000
## 1405	1.3	13.0	1.13	2.82	2800	130000
## 1406	4.9	11.0	1.99	4.98	78	110000
## 1407	26.0	30.0	-	-	94000	210000
## 1408	38.0	42.0	-	-	90000	220000
## 1409	30.0	39.0	-	-	72000	280000
## 1410	3.0	29.0	0.32	1.22	13	3300000
## 1411	1.0	3.8	0.32	1.13	23	130000
## 1412	1.0	2.6	0.32	2.96	17	32
## 1413	1.1	2.4	0.32	1.02	22	47
## 1414	1.0	4.0	0.32	0.83	22	49
## 1415	1.1	3.2	0.32	1.22	21	40
## 1416	1.0	2.9	0.32	1.02	22	40
## 1417	2.4	2.8	0.02	1.82	33	64
## 1418	2.2	2.6	0.02	0.32	26	48
## 1419	2.0	2.0	0.80	4.10	40	1600

## 1420	2.0	2.0	1.40	4.90	17	1600
## 1421	2.0	3.0	1.10	3.90	33	1600
## 1422	2.0	2.0	0.60	4.10	21	1600
## 1423	1.0	2.2	0.30	0.70	84	120
## 1424	1.0	2.3	0.30	0.80	84	110
## 1425	1.0	2.2	0.30	0.80	79	150
## 1426	1.0	2.3	0.30	0.70	100	940
## 1427	1.0	1.8	0.30	0.30	31	120
## 1428	1.0	1.4	0.30	0.30	31	180
## 1429	1.0	1.9	0.30	0.90	200	310
## 1430	1.0	2.6	0.30	2.30	200	2000
## 1431	1.0	2.1	0.30	2.70	2	400
## 1432	1.0	3.3	0.30	0.30	17	4800
## 1433	1.0	1.9	0.30	0.31	100	7000
## 1434	1.0	1.0	0.30	0.38	100	7000
## 1435	1.0	1.0	0.30	0.30	600	3400
## 1436	1.0	3.2	0.30	3.90	1100	4800
## 1437	1.0	1.9	0.30	0.60	400	5800
## 1438	1.2	2.4	0.30	1.10	70	210
## 1439	1.0	2.1	0.30	0.60	79	140
## 1440	1.0	3.3	0.30	0.73	27	94
## 1441	2.0	4.0	0.30	0.70	270	2800
## 1442	1.0	3.6	0.30	0.60	200	4000
## 1443	1.6	2.6	0.30	0.70	94	140
## 1444	1.0	3.3	0.30	0.80	1000	70000
## 1445	1.0	3.3	0.30	3.70	2	920
## 1446	1.0	1.6	0.30	0.30	100	200
## 1447	4.1	8.2	1.10	1.50	4000	4900
## 1448	1.0	1.7	0.30	0.30	32	110
## 1449	1.0	2.1	0.30	1.00	200	1000
## 1450	1.0	2.1	0.30	0.30	20	100
## 1451	1.1	2.4	0.30	0.80	200	490
## 1452	1.0	3.5	0.30	3.30	2	400
## 1453	1.0	1.5	0.30	0.30	20	84
## 1454	1.0	2.2	0.30	0.40	20	120
## 1455	1.0	1.6	0.30	0.30	84	2500
## 1456	1.0	1.8	0.30	0.30	100	17000
## 1457	1.0	2.5	0.30	0.30	200	400
## 1458	1.0	2.4	0.30	0.30	2	2200
## 1459	1.0	3.1	0.30	0.30	2	1200
## 1460	1.0	1.8	0.30	0.40	2	70
## 1461	1.0	2.0	0.30	0.30	17	150
## 1462	1.0	1.0	0.30	2.02	100	1100
## 1463	1.0	2.9	0.30	0.70	27	250
## 1464	1.0	1.3	0.30	1.00	2	1500
## 1465	1.0	2.3	0.30	0.40	39	120
## 1466	1.0	2.1	0.30	0.60	58	150
## 1467	1.0	2.3	0.30	0.30	100	200
## 1468	1.0	2.5	0.30	0.30	100	310
## 1469	1.8	2.2	0.30	0.80	100	140
## 1470	1.0	1.0	0.30	0.70	100	170
## 1471	1.0	1.5	0.30	1.20	61	170
## 1472	1.0	1.0	0.30	0.42	110	680
## 1473	1.0	1.7	0.30	0.90	91	210

## 1474	1.0	1.0	0.30	0.30	80	270
## 1475	1.0	1.6	0.30	0.34	220	2000
## 1476	1.0	3.6	0.30	0.30	200	4800
## 1477	1.0	2.3	0.30	0.30	63	3400
## 1478	1.0	2.9	0.30	0.30	310	1400
## 1479	1.0	1.7	0.30	1.30	310	6300
## 1480	1.0	3.9	0.30	0.40	70	1300
## 1481	1.0	2.7	0.30	0.30	340	7000
## 1482	1.0	3.3	0.30	2.60	94	4000
## 1483	1.0	2.1	0.30	0.30	20	84
## 1484	1.0	1.8	0.30	0.31	24	120
## 1485	1.0	2.2	0.30	0.60	310	1500
## 1486	1.0	2.1	0.30	0.40	40	84
## 1487	1.0	1.7	0.30	0.30	27	94
## 1488	1.0	2.0	0.30	0.30	31	100
## 1489	1.0	2.4	0.30	1.20	36	79
## 1490	1.0	2.9	0.30	0.40	40	79
## 1491	1.0	2.4	0.30	0.60	48	210
## 1492	1.0	2.1	0.30	0.30	20	94
## 1493	1.0	2.0	0.30	0.30	100	310
## 1494	3.0	4.0	0.30	0.40	2	110
## 1495	3.0	4.0	0.30	0.51	2	23
## 1496	3.0	4.0	0.30	0.60	5	27
## 1497	3.0	4.0	0.30	0.51	2	21
## 1498	3.0	8.0	0.30	1.60	2	26
## 1499	2.4	3.2	0.90	4.27	4	11
## 1500	3.0	9.0	0.30	3.10	2	79
## 1501	3.0	12.0	0.30	1.20	2	33
## 1502	3.2	4.0	0.30	0.89	2	49
## 1503	3.6	17.0	0.30	2.40	2	170
## 1504	3.2	45.0	0.30	15.90	5	17000
## 1505	1.4	1.8	0.30	0.99	4	6
## 1506	3.0	10.0	0.30	9.20	2	48
## 1507	3.0	3.8	0.30	2.30	5	140
## 1508	3.2	7.4	0.30	5.90	2	33
## 1509	3.0	7.4	0.30	6.20	2	140
## 1510	3.2	3.8	0.30	0.60	2	240
## 1511	3.0	7.4	0.30	3.50	5	79
## 1512	3.2	3.8	0.30	0.30	2	70
## 1513	3.0	4.0	0.30	1.12	2	22
## 1514	1.6	2.0	0.30	2.35	5	14
## 1515	1.4	2.8	0.30	4.20	2	17
## 1516	1.4	3.8	0.30	4.72	6	14
## 1517	1.4	2.2	0.30	2.27	2	6
## 1518	1.4	2.0	0.30	1.26	2	6
## 1519	3.2	3.8	0.30	0.55	5	27
## 1520	3.0	3.8	0.30	1.00	5	33
## 1521	3.0	3.8	0.30	1.20	2	110
## 1522	3.0	4.0	0.30	0.79	5	94
## 1523	3.2	3.8	0.30	0.51	5	17
## 1524	1.6	1.8	0.30	2.25	4	5
## 1525	1.2	1.8	0.30	0.80	4	6
## 1526	1.4	1.8	0.30	2.19	4	5
## 1527	2.9	3.3	<NA>	<NA>	5	20

## 1528	2.9	3.3	0.32	0.32	50	95
## 1529	3.0	3.5	-	-	40	85
## 1530	2.8	3.3	-	-	35	60
## 1531	3.1	5.1	-	-	10	180
## 1532	3.0	3.5	-	-	55	90
## 1533	3.0	4.9	-	-	50	195
## 1534	3.0	3.4	-	-	35	95
## 1535	2.9	3.0	-	-	5	5
## 1536	2.8	3.2	-	-	25	45
## 1537	3.0	5.3	-	-	60	285
## 1538	3.0	3.6	-	-	50	95
## 1539	2.9	3.4	-	-	35	60
## 1540	3.0	4.5	-	-	40	80
## 1541	3.0	3.4	-	-	45	80
## 1542	2.8	3.5	-	-	2	25
## 1543	3.1	3.5	-	-	15	200
## 1544	2.0	3.5	-	-	30	70
## 1545	2.8	3.4	-	-	10	35
## 1546	3.0	3.3	-	-	50	80
## 1547	3.0	3.7	-	-	50	85
## 1548	3.2	3.6	-	-	50	90
## 1549	3.9	6.2	-	-	215	295
## 1550	3.8	5.9	-	-	200	295
## 1551	3.3	4.7	-	-	80	165
## 1552	3.3	4.8	0.30	0.30	95	190
## 1553	3.0	3.8	-	-	40	75
## 1554	2.8	3.0	-	-	5	25
## 1555	2.7	3.4	-	-	15	35
## 1556	3.0	3.1	-	-	20	30
## 1557	3.0	3.5	-	-	15	30
## 1558	3.1	3.3	-	-	40	65
## 1559	2.8	4.1	-	-	25	175
## 1560	2.9	3.0	-	-	10	30
## 1561	2.9	3.6	-	-	40	65
## 1562	3.1	3.2	-	-	45	85
## 1563	1.4	1.8	0.42	0.52	21	36
## 1564	1.4	1.8	0.32	0.56	31	43
## 1565	1.3	1.7	0.37	0.52	130	170
## 1566	1.4	1.8	0.44	0.55	23	49
## 1567	1.8	2.6	0.32	0.32	2	2
## 1568	1.3	1.6	0.37	0.55	20	31
## 1569	1.0	1.6	0.32	0.36	27	38
## 1570	2.1	2.9	0.32	0.62	300	430
## 1571	1.2	1.8	0.32	1.22	17	27
## 1572	1.3	2.0	0.34	1.22	31	63
## 1573	1.0	1.7	0.32	1.32	2	4
## 1574	1.3	2.0	0.32	0.82	13	31
## 1575	1.0	1.7	0.37	0.72	6	17
## 1576	1.1	1.7	0.32	0.36	2	8
## 1577	1.5	2.2	2.02	3.42	17	34
## 1578	14.0	35.0	0.39	4.16	6300	17000
## 1579	15.6	45.0	0.33	5.00	14000	32000
## 1580	15.1	40.0	0.40	1.90	13000	30000
## 1581	6.2	14.2	1.12	5.16	4200	11000

## 1582	16.0	39.0	0.32	2.60	13000	27000
## 1583	1.4	1.8	0.32	0.52	2	17
## 1584	6.0	24.0	0.32	5.57	4100	8400
## 1585	2.3	2.7	0.32	0.52	210	380
## 1586	1.0	1.7	0.32	0.47	43	63
## 1587	1.0	2.3	0.02	0.12	3	1100
## 1588	1.0	1.7	0.02	0.11	2	240
## 1589	1.1	2.4	0.02	0.10	-	-
## 1590	1.0	2.4	0.02	0.02	3	42
## 1591	1.0	1.7	0.02	0.04	2	95
## 1592	1.0	2.4	0.02	0.21	10	1100
## 1593	1.0	2.0	0.02	0.10	3	1100
## 1594	1.0	1.8	0.02	0.02	-	-
## 1595	1.0	2.6	0.02	0.06	3	1100
## 1596	1.0	1.9	0.02	0.05	-	-
## 1597	1.0	2.2	0.02	25.20	-	-
## 1598	1.0	1.6	0.02	0.02	-	-
## 1599	1.0	2.1	0.02	0.05	3	1100
## 1600	1.0	2.5	0.37	2.02	330	2300
## 1601	1.0	2.2	0.32	1.02	78	1700
## 1602	1.0	2.1	0.32	3.39	330	2300
## 1603	1.0	1.3	0.32	1.02	20	1700
## 1604	1.0	1.9	0.32	1.22	78	2200
## 1605	1.0	1.5	0.32	1.52	78	1400
## 1606	1.0	1.6	0.32	1.22	20	1300
## 1607	1.0	2.0	0.32	1.32	130	2200
## 1608	1.0	2.1	0.32	1.45	2	490
## 1609	1.0	2.6	0.32	1.09	45	2200
## 1610	1.0	1.8	0.32	1.22	130	2200
## 1611	1.0	1.8	0.32	1.52	130	2400
## 1612	1.0	1.2	0.32	1.02	130	1700
## 1613	1.0	1.3	0.32	1.42	20	1300
## 1614	1.0	1.3	0.32	0.92	2	300
## 1615	20.0	20.0	0.32	0.32	-	-
## 1616	25.0	25.0	0.32	0.32	-	-
## 1617	2.0	2.0	0.32	0.32	-	-
## 1618	1.0	1.0	0.32	0.32	-	-
## 1619	14.0	47.0	1.10	3.80	3900	4900
## 1620	11.0	37.0	1.00	2.50	2700	4600
## 1621	13.0	43.0	1.00	1.60	3200	6300
## 1622	14.0	27.0	1.10	1.40	3400	6300
## 1623	14.0	38.0	1.10	1.50	3400	4900
## 1624	9.0	23.0	1.00	2.70	2400	3400
## 1625	14.0	41.0	1.20	3.10	3900	6300
## 1626	6.0	25.0	0.32	1.88	2100	1700000
## 1627	7.0	24.0	0.90	1.70	1700	4100
## 1628	12.0	42.0	1.00	2.90	3200	4900
## 1629	4.0	16.0	2.91	3.64	78	2600
## 1630	5.0	25.0	0.32	0.66	4900	49000
## 1631	10.0	22.0	1.00	1.20	2600	4000
## 1632	11.0	33.0	1.00	1.40	2600	4700
## 1633	6.0	19.0	0.30	2.50	1700	2700
## 1634	13.0	40.0	1.00	1.40	2600	6300
## 1635	12.0	43.0	1.00	2.90	3300	4700

## 1636	1.0	4.0	0.32	0.62	2	460
## 1637	2.0	7.0	0.32	1.42	31	110
## 1638	1.0	3.0	0.19	0.45	2	2
## 1639	1.0	3.0	0.18	0.66	2	20
## 1640	2.0	7.5	0.32	6.42	26	130
## 1641	1.0	3.0	0.32	0.90	2	17
## 1642	2.0	3.0	0.32	0.56	17	70
## 1643	1.0	2.6	0.32	0.82	2	13
## 1644	1.0	3.0	0.32	0.69	2	8
## 1645	2.0	3.0	0.17	0.57	11	94
## 1646	2.0	3.0	0.32	0.84	21	84
## 1647	2.0	3.0	0.19	0.45	12	70
## 1648	2.0	3.2	0.32	5.15	4	63
## 1649	2.0	3.0	0.32	0.80	17	110
## 1650	1.0	2.3	0.92	2.90	17	63
## 1651	1.1	1.9	0.80	2.40	21	63
## 1652	1.2	2.1	0.88	2.59	33	94
## 1653	1.5	2.9	0.68	6.76	47	120
## 1654	1.3	2.2	1.30	3.20	33	84
## 1655	1.5	2.4	1.08	3.40	41	94
## 1656	1.1	2.1	0.68	2.34	26	63
## 1657	1.0	1.8	0.58	2.50	38	84
## 1658	1.6	2.8	0.85	4.30	58	94
## 1659	1.5	3.2	2.10	4.40	63	110
## 1660	1.2	2.5	0.65	3.82	15	63
## 1661	1.4	2.8	1.80	4.40	33	63
## 1662	1.2	2.1	0.81	2.48	26	58
## 1663	1.1	2.2	0.92	3.90	31	63
## 1664	1.1	2.2	0.80	3.10	25	63
## 1665	1.0	1.9	0.90	3.00	27	63
##	TotalColiformMin	TotalColiformMax	year			
## 1		150	640	2017		
## 2		120	460	2017		
## 3		17	540	2017		
## 4		110	1600	2017		
## 5		240	1600	2017		
## 6		33	1600	2017		
## 7		70	1600	2017		
## 8		110	1600	2017		
## 9		120	1600	2017		
## 10		33	1600	2017		
## 11		63	1600	2017		
## 12		450	28000	2017		
## 13		78	132000	2017		
## 14		780	7900	2017		
## 15		330	13000	2017		
## 16		780	7900	2017		
## 17		4600	54000	2017		
## 18		230	5400	2017		
## 19		110	9200	2017		
## 20		46	9200	2017		
## 21		1300	9200	2017		
## 22		460	5400	2017		
## 23		1300	24000	2017		

## 24	490	17000 2017
## 25	1100	9400 2017
## 26	2700	92000 2017
## 27	780	13000 2017
## 28	7900	35000 2017
## 29	330	17000 2017
## 30	170	1700 2017
## 31	17000	92000 2017
## 32	170	3300 2017
## 33	310	22000 2017
## 34	170	11000 2017
## 35	2	177 2017
## 36	94	1600 2017
## 37	2	210 2017
## 38	2	540 2017
## 39	700	17000 2017
## 40	13	140 2017
## 41	110	240 2017
## 42	276	542 2017
## 43	212	542 2017
## 44	175	542 2017
## 45	2	542 2017
## 46	3	210 2017
## 47	170	700 2017
## 48	50	110 2017
## 49	14	22 2017
## 50	240	2400 2017
## 51	11	47 2017
## 52	<NA>	<NA> 2017
## 53	<NA>	<NA> 2017
## 54	27800	54200 2017
## 55	1090	9400 2017
## 56	<NA>	<NA> 2017
## 57	24	39 2017
## 58	21	32 2017
## 59	26	79 2017
## 60	24	34 2017
## 61	46	920 2017
## 62	25	33 2017
## 63	140	1600 2017
## 64	27	540 2017
## 65	79	920 2017
## 66	130	1800 2017
## 67	79	1600 2017
## 68	94	1600 2017
## 69	1380	7900 2017
## 70	1160	17000 2017
## 71	20	1280 2017
## 72	40	1680 2017
## 73	310	500 2017
## 74	1200	4000 2017
## 75	140	790000 2017
## 76	38200	41000 2017
## 77	240	480 2017

## 78	600	1400000 2017
## 79	1800	3800 2017
## 80	72	460 2017
## 81	260	840 2017
## 82	260	560 2017
## 83	180	320 2017
## 84	620	1750 2017
## 85	140	460 2017
## 86	550	880 2017
## 87	320	400 2017
## 88	250	680 2017
## 89	32	480 2017
## 90	20	400 2017
## 91	18	1400 2017
## 92	49	3000 2017
## 93	91	1180 2017
## 94	34	1230 2017
## 95	84	990 2017
## 96	220	7900 2017
## 97	200	400 2017
## 98	1380	1600 2017
## 99	1000	1330 2017
## 100	360	480 2017
## 101	320	420 2017
## 102	100	700 2017
## 103	210	700 2017
## 104	190	750 2017
## 105	12	240 2017
## 106	24	2400 2017
## 107	70	14000 2017
## 108	100	4800 2017
## 109	380	2800 2017
## 110	350	1740 2017
## 111	330	1020 2017
## 112	140	1700 2017
## 113	130	1720 2017
## 114	120	2600 2017
## 115	89	1200 2017
## 116	51	1120 2017
## 117	66	1520 2017
## 118	62	12000 2017
## 119	48	1760 2017
## 120	72	1240 2017
## 121	480	1280 2017
## 122	20	200 2017
## 123	40	320 2017
## 124	60	980 2017
## 125	700	7900 2017
## 126	430	7900 2017
## 127	1060	7900 2017
## 128	120	79000 2017
## 129	780	10000 2017
## 130	350	9400 2017
## 131	40	180 2017

## 132	82	800 2017
## 133	12	540 2017
## 134	4	580 2017
## 135	200	300 2017
## 136	260	630 2017
## 137	200	14000 2017
## 138	90	140 2017
## 139	20	1430 2017
## 140	110	540 2017
## 141	5	540 2017
## 142	17	220 2017
## 143	24	170 2017
## 144	52	140 2017
## 145	8	130 2017
## 146	33	140 2017
## 147	5	210 2017
## 148	70	94 2017
## 149	920	1600 2017
## 150	43	280 2017
## 151	31	430 2017
## 152	15	220 2017
## 153	23	140 2017
## 154	23	170 2017
## 155	110	170 2017
## 156	70	280 2017
## 157	46	280 2017
## 158	46	170 2017
## 159	79	240 2017
## 160	22	110 2017
## 161	34	130 2017
## 162	49	280 2017
## 163	8	540 2017
## 164	9	110 2017
## 165	58	280 2017
## 166	94	430 2017
## 167	63	350 2017
## 168	43	540 2017
## 169	6	70 2017
## 170	22	79 2017
## 171	20	140 2017
## 172	25	94 2017
## 173	21	79 2017
## 174	33	150 2017
## 175	26	140 2017
## 176	175	190 2017
## 177	60	80 2017
## 178	150	170 2017
## 179	120	130 2017
## 180	250	9500 2017
## 181	3700	7200 2017
## 182	115	250 2017
## 183	120	270 2017
## 184	50	82 2017
## 185	80	93 2017

## 186	85	145 2017
## 187	75	130 2017
## 188	40	60 2017
## 189	70	75 2017
## 190	80	87 2017
## 191	3	220 2017
## 192	120	130 2017
## 193	120	350 2017
## 194	75	350 2017
## 195	69	80 2017
## 196	70	73 2017
## 197	40	59 2017
## 198	30	87 2017
## 199	30	39 2017
## 200	29	36 2017
## 201	10	25 2017
## 202	130	230 2017
## 203	11	25 2017
## 204	80	190 2017
## 205	35	120 2017
## 206	45	110 2017
## 207	80	150 2017
## 208	60	130 2017
## 209	220	740 2017
## 210	17	3800 2017
## 211	59	3500 2017
## 212	210	590 2017
## 213	110	670 2017
## 214	40	94 2017
## 215	39	79 2017
## 216	38	74 2017
## 217	33	110 2017
## 218	47	79 2017
## 219	27	70 2017
## 220	35	94 2017
## 221	31	79 2017
## 222	25	79 2017
## 223	22000	54000 2017
## 224	7900	54000 2017
## 225	54000	95000 2017
## 226	49000	123000 2017
## 227	31	120 2017
## 228	33000	93000 2017
## 229	14000	52000 2017
## 230	50	240 2017
## 231	2	50 2017
## 232	94	350 2017
## 233	19	1200 2017
## 234	30	123 2017
## 235	24	110 2017
## 236	<NA>	<NA> 2017
## 237	<NA>	<NA> 2017
## 238	<NA>	<NA> 2017
## 239	<NA>	<NA> 2017

## 240	11000	28000 2017
## 241	15000	70000 2017
## 242	17000	35000 2017
## 243	14000	28000 2017
## 244	14000	24000 2017
## 245	11000	22000 2017
## 246	13000	24000 2017
## 247	17000	33000 2017
## 248	14000	28000 2017
## 249	14000	35000 2017
## 250	13000	28000 2017
## 251	11000	26000 2017
## 252	38000	470000 2017
## 253	7	28 2017
## 254	7	28 2017
## 255	7	75 2017
## 256	40	280 2017
## 257	34	170 2017
## 258	50	220 2017
## 259	50	220 2017
## 260	33	140 2017
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## 1076	-	- 2020
## 1077	-	- 2020
## 1078	-	- 2020
## 1079	1200	1600 2020
## 1080	1200	1600 2020
## 1081	141	473 2020
## 1082	2	1600 2020
## 1083	2	1600 2020
## 1084	120	432 2020
## 1085	2	1600 2020
## 1086	2	1600 2020
## 1087	14	70 2020
## 1088	20	76 2020
## 1089	32	63 2020
## 1090	26	72 2020
## 1091	11	67 2020
## 1092	20	84 2020
## 1093	540	5400 2020
## 1094	920	3500 2020
## 1095	400	7900 2020
## 1096	630	6300 2020
## 1097	140	7000 2020
## 1098	61	8400 2020
## 1099	180	350 2020
## 1100	840	4300 2020
## 1101	130	2600 2020
## 1102	2600	17000 2020
## 1103	280	920 2020

## 1104	400	63000 2020
## 1105	490	3600 2020
## 1106	200	1700 2020
## 1107	180	450 2020
## 1108	280	580 2020
## 1109	200	1500 2020
## 1110	200	630 2020
## 1111	200	480 2020
## 1112	400	790 2020
## 1113	120	320 2020
## 1114	220	700 2020
## 1115	17	240 2020
## 1116	110	4900 2020
## 1117	220	18000 2020
## 1118	150	560 2020
## 1119	280	490 2020
## 1120	60	280 2020
## 1121	240	380 2020
## 1122	40	7900 2020
## 1123	300	600 2020
## 1124	480	920 2020
## 1125	310	700 2020
## 1126	210	380 2020
## 1127	180	390 2020
## 1128	200	3500 2020
## 1129	200	540 2020
## 1130	200	2200 2020
## 1131	20	280 2020
## 1132	12	200 2020
## 1133	700	46000 2020
## 1134	170	4000 2020
## 1135	170	480 2020
## 1136	170	600 2020
## 1137	140	350 2020
## 1138	210	680 2020
## 1139	170	430 2020
## 1140	150	810 2020
## 1141	200	450 2020
## 1142	130	400 2020
## 1143	94	430 2020
## 1144	200	390 2020
## 1145	94	410 2020
## 1146	140	360 2020
## 1147	170	600 2020
## 1148	2	1700 2020
## 1149	79	6300 2020
## 1150	150	1400 2020
## 1151	580	3500 2020
## 1152	120	84000 2020
## 1153	1100	5400 2020
## 1154	140	4800 2020
## 1155	63	7900 2020
## 1156	1000	4900 2020
## 1157	20	170 2020

## 1158	12	180 2020
## 1159	170	3100 2020
## 1160	310	7900 2020
## 1161	220	460 2020
## 1162	350	840 2020
## 1163	63	2400 2020
## 1164	310	630 2020
## 1165	180	490 2020
## 1166	430	8100 2020
## 1167	610	20000 2020
## 1168	4	26 2020
## 1169	32	48 2020
## 1170	43	49 2020
## 1171	47	63 2020
## 1172	9	1600 2020
## 1173	7	49 2020
## 1174	39	63 2020
## 1175	34	49 2020
## 1176	31	63 2020
## 1177	11	43 2020
## 1178	17	49 2020
## 1179	22	920 2020
## 1180	23	540 2020
## 1181	21	540 2020
## 1182	26	920 2020
## 1183	22	280 2020
## 1184	8	280 2020
## 1185	23	540 2020
## 1186	17	350 2020
## 1187	11	48 2020
## 1188	920	160000 2020
## 1189	2	130 2020
## 1190	21	140 2020
## 1191	21	540 2020
## 1192	17	540 2020
## 1193	17	540 2020
## 1194	79	140 2020
## 1195	2	350 2020
## 1196	22	540 2020
## 1197	12	48 2020
## 1198	2	430 2020
## 1199	13	540 2020
## 1200	17	920 2020
## 1201	17	1600 2020
## 1202	17	920 2020
## 1203	12	540 2020
## 1204	12	94 2020
## 1205	17	94 2020
## 1206	20	170 2020
## 1207	17	150 2020
## 1208	2	240 2020
## 1209	2	220 2020
## 1210	2	79 2020
## 1211	23	39 2020

## 1212	33	140 2020
## 1213	12	58 2020
## 1214	14	41 2020
## 1215	50	190 2020
## 1216	85	195 2020
## 1217	75	180 2020
## 1218	80	225 2020
## 1219	470	1135 2020
## 1220	470	1580 2020
## 1221	120	190 2020
## 1222	105	185 2020
## 1223	130	175 2020
## 1224	145	335 2020
## 1225	85	155 2020
## 1226	95	210 2020
## 1227	60	170 2020
## 1228	145	235 2020
## 1229	65	260 2020
## 1230	335	1015 2020
## 1231	10	20 2020
## 1232	235	360 2020
## 1233	95	165 2020
## 1234	80	1675 2020
## 1235	50	260 2020
## 1236	70	280 2020
## 1237	35	120 2020
## 1238	10	30 2020
## 1239	30	55 2020
## 1240	110	221 2020
## 1241	95	250 2020
## 1242	10	55 2020
## 1243	45	175 2020
## 1244	140	275 2020
## 1245	130	165 2020
## 1246	790	1300 2020
## 1247	2	17 2020
## 1248	410	930 2020
## 1249	210	430 2020
## 1250	280	470 2020
## 1251	140	430 2020
## 1252	58	94 2020
## 1253	79	140 2020
## 1254	220	410 2020
## 1255	120	280 2020
## 1256	230	350 2020
## 1257	49	120 2020
## 1258	120	92000 2020
## 1259	31	94 2020
## 1260	9200	70000 2020
## 1261	7900	38000 2020
## 1262	21000	120000 2020
## 1263	17000	110000 2020
## 1264	49	140 2020
## 1265	25000	79000 2020

## 1266	5800	39000 2020
## 1267	240	840 2020
## 1268	2	6 2020
## 1269	21	170 2020
## 1270	110	140 2020
## 1271	84	210 2020
## 1272	70	240 2020
## 1273	210	230 2020
## 1274	140	330 2020
## 1275	110	330 2020
## 1276	-	- 2020
## 1277	-	- 2020
## 1278	-	- 2020
## 1279	-	- 2020
## 1280	11000	28000 2020
## 1281	17000	35000 2020
## 1282	22000	35000 2020
## 1283	21000	35000 2020
## 1284	17000	35000 2020
## 1285	12000	28000 2020
## 1286	15000	35000 2020
## 1287	21000	35000 2020
## 1288	17000	28000 2020
## 1289	17000	43000 2020
## 1290	17000	35000 2020
## 1291	14000	28000 2020
## 1292	21000	35000 2020
## 1293	22000	43000 2020
## 1294	280000	430000 2020
## 1295	24000	9200000 2020
## 1296	160000	240000 2020
## 1297	2	1700 2020
## 1298	28	210 2020
## 1299	150	240 2020
## 1300	39	120 2020
## 1301	64	120 2020
## 1302	2	33 2020
## 1303	10	84 2020
## 1304	34	430 2020
## 1305	70	430 2020
## 1306	4	43 2020
## 1307	2	91 2020
## 1308	110	430 2020
## 1309	2	1200 2020
## 1310	94	350 2020
## 1311	2	61 2020
## 1312	110	220 2020
## 1313	46	280 2020
## 1314	58	350 2020
## 1315	70	350 2020
## 1316	84	220 2020
## 1317	130	280 2020
## 1318	94	240 2020
## 1319	170	430 2020

## 1320	58	240 2020
## 1321	58	210 2020
## 1322	58	210 2020
## 1323	84	240 2020
## 1324	70	220 2020
## 1325	70	170 2020
## 1326	63	130 2020
## 1327	110	220 2020
## 1328	70	170 2020
## 1329	63	170 2020
## 1330	70	170 2020
## 1331	63	150 2020
## 1332	63	240 2020
## 1333	40	210 2020
## 1334	63	240 2020
## 1335	46	220 2020
## 1336	63	280 2020
## 1337	84	240 2020
## 1338	58	210 2020
## 1339	58	170 2020
## 1340	58	1110 2020
## 1341	63	210 2020
## 1342	Min	Max 2021
## 1343	93	150 2021
## 1344	150	210 2021
## 1345	120	120 2021
## 1346	150	210 2021
## 1347	120	210 2021
## 1348	120	150 2021
## 1349	24	39 2021
## 1350	119	346 2021
## 1351	297	427 2021
## 1352	74	91 2021
## 1353	31	38 2021
## 1354	218	538 2021
## 1355	520	1700 2021
## 1356	710	1700 2021
## 1357	1600	28000 2021
## 1358	14	69 2021
## 1359	4900	14000 2021
## 1360	390	2800 2021
## 1361	490	3300 2021
## 1362	790	3500 2021
## 1363	790	3500 2021
## 1364	270	1700 2021
## 1365	790	3300 2021
## 1366	140	2200 2021
## 1367	230	3500 2021
## 1368	13	490 2021
## 1369	110	700 2021
## 1370	170	700 2021
## 1371	490	1700 2021
## 1372	130	460 2021
## 1373	330	9400 2021

## 1374	7000	54000 2021
## 1375	78	2800 2021
## 1376	790	3200 2021
## 1377	490	4900 2021
## 1378	8	790 2021
## 1379	270	1400 2021
## 1380	330	5400 2021
## 1381	230	3900 2021
## 1382	2700	7900 2021
## 1383	1100	5400 2021
## 1384	460	3300 2021
## 1385	110	1600 2021
## 1386	40	120 2021
## 1387	350	1600 2021
## 1388	280	1600 2021
## 1389	210	1600 2021
## 1390	39	540 2021
## 1391	8	140 2021
## 1392	49	17000 2021
## 1393	94	1600 2021
## 1394	26	49 2021
## 1395	1600	1600 2021
## 1396	34	63 2021
## 1397	84	1600 2021
## 1398	34	84 2021
## 1399	1600	1600 2021
## 1400	350	1600 2021
## 1401	2	9 2021
## 1402	2300	23000 2021
## 1403	200	42600 2021
## 1404	300	148000 2021
## 1405	16000	270000 2021
## 1406	490	210000 2021
## 1407	220000	390000 2021
## 1408	210000	320000 2021
## 1409	150000	420000 2021
## 1410	23	4900000 2021
## 1411	23	170000 2021
## 1412	70	170 2021
## 1413	110	430 2021
## 1414	94	350 2021
## 1415	79	350 2021
## 1416	110	350 2021
## 1417	109	176 2021
## 1418	94	141 2021
## 1419	130	1600 2021
## 1420	160	1600 2021
## 1421	280	1600 2021
## 1422	160	1600 2021
## 1423	350	470 2021
## 1424	320	390 2021
## 1425	350	490 2021
## 1426	390	2500 2021
## 1427	140	580 2021

## 1428	130	490 2021
## 1429	490	790 2021
## 1430	400	2500 2021
## 1431	49	1100 2021
## 1432	94	7900 2021
## 1433	310	9400 2021
## 1434	400	8400 2021
## 1435	840	4700 2021
## 1436	1700	7900 2021
## 1437	490	7000 2021
## 1438	350	540 2021
## 1439	340	460 2021
## 1440	63	200 2021
## 1441	540	6300 2021
## 1442	480	7900 2021
## 1443	350	580 2021
## 1444	2000	84000 2021
## 1445	100	2200 2021
## 1446	400	610 2021
## 1447	17000	21000 2021
## 1448	92	380 2021
## 1449	490	2500 2021
## 1450	140	430 2021
## 1451	490	920 2021
## 1452	100	1500 2021
## 1453	150	430 2021
## 1454	150	540 2021
## 1455	110	3100 2021
## 1456	400	25000 2021
## 1457	580	840 2021
## 1458	2	5800 2021
## 1459	2	2700 2021
## 1460	26	140 2021
## 1461	40	480 2021
## 1462	350	2800 2021
## 1463	170	700 2021
## 1464	100	3100 2021
## 1465	220	600 2021
## 1466	310	630 2021
## 1467	310	630 2021
## 1468	400	630 2021
## 1469	400	580 2021
## 1470	310	580 2021
## 1471	200	480 2021
## 1472	340	1320 2021
## 1473	270	630 2021
## 1474	260	630 2021
## 1475	790	4800 2021
## 1476	480	6300 2021
## 1477	110	6300 2021
## 1478	700	4900 2021
## 1479	790	7000 2021
## 1480	120	2500 2021
## 1481	840	40000 2021

## 1482	540	7000 2021
## 1483	170	430 2021
## 1484	120	600 2021
## 1485	700	3100 2021
## 1486	170	540 2021
## 1487	100	630 2021
## 1488	170	380 2021
## 1489	70	170 2021
## 1490	82	200 2021
## 1491	170	700 2021
## 1492	200	430 2021
## 1493	400	700 2021
## 1494	14	350 2021
## 1495	22	130 2021
## 1496	22	220 2021
## 1497	17	94 2021
## 1498	21	110 2021
## 1499	47	70 2021
## 1500	17	280 2021
## 1501	11	220 2021
## 1502	17	540 2021
## 1503	8	540 2021
## 1504	27	92000 2021
## 1505	11	25 2021
## 1506	17	920 2021
## 1507	22	1600 2021
## 1508	11	280 2021
## 1509	22	540 2021
## 1510	17	1600 2021
## 1511	17	540 2021
## 1512	17	540 2021
## 1513	11	140 2021
## 1514	17	47 2021
## 1515	2	49 2021
## 1516	24	48 2021
## 1517	2	39 2021
## 1518	2	40 2021
## 1519	17	79 2021
## 1520	22	170 2021
## 1521	17	220 2021
## 1522	17	210 2021
## 1523	17	70 2021
## 1524	12	27 2021
## 1525	13	39 2021
## 1526	11	25 2021
## 1527	40	50 2021
## 1528	160	250 2021
## 1529	140	270 2021
## 1530	85	180 2021
## 1531	25	945 2021
## 1532	140	275 2021
## 1533	130	450 2021
## 1534	100	210 2021
## 1535	30	45 2021

## 1536	60	95 2021
## 1537	120	990 2021
## 1538	130	215 2021
## 1539	85	140 2021
## 1540	125	195 2021
## 1541	170	250 2021
## 1542	5	50 2021
## 1543	240	515 2021
## 1544	100	195 2021
## 1545	50	75 2021
## 1546	125	190 2021
## 1547	165	210 2021
## 1548	145	230 2021
## 1549	775	1115 2021
## 1550	735	1015 2021
## 1551	160	295 2021
## 1552	215	885 2021
## 1553	75	180 2021
## 1554	15	70 2021
## 1555	35	270 2021
## 1556	65	100 2021
## 1557	20	75 2021
## 1558	155	195 2021
## 1559	60	300 2021
## 1560	60	80 2021
## 1561	115	170 2021
## 1562	130	215 2021
## 1563	280	470 2021
## 1564	230	350 2021
## 1565	340	460 2021
## 1566	280	440 2021
## 1567	2	24 2021
## 1568	140	170 2021
## 1569	280	440 2021
## 1570	920	1700 2021
## 1571	310	430 2021
## 1572	350	700 2021
## 1573	8	94 2021
## 1574	240	540 2021
## 1575	79	140 2021
## 1576	84	140 2021
## 1577	310	630 2021
## 1578	21000	43000 2021
## 1579	49000	110000 2021
## 1580	46000	94000 2021
## 1581	12000	24000 2021
## 1582	48000	94000 2021
## 1583	49	120 2021
## 1584	9400	17000 2021
## 1585	1300	2300 2021
## 1586	410	700 2021
## 1587	120	2400 2021
## 1588	75	1100 2021
## 1589	-	- 2021

## 1590	93	1100 2021
## 1591	210	1100 2021
## 1592	120	2400 2021
## 1593	53	2400 2021
## 1594	-	- 2021
## 1595	20	2400 2021
## 1596	-	- 2021
## 1597	-	- 2021
## 1598	-	- 2021
## 1599	20	2400 2021
## 1600	1700	4900 2021
## 1601	330	3500 2021
## 1602	940	4900 2021
## 1603	78	3500 2021
## 1604	330	4700 2021
## 1605	220	3500 2021
## 1606	110	3500 2021
## 1607	1100	3500 2021
## 1608	2	3500 2021
## 1609	230	4000 2021
## 1610	790	4900 2021
## 1611	790	4700 2021
## 1612	460	4000 2021
## 1613	45	3500 2021
## 1614	280	1600 2021
## 1615	-	- 2021
## 1616	-	- 2021
## 1617	-	- 2021
## 1618	-	- 2021
## 1619	24000	35000 2021
## 1620	21000	28000 2021
## 1621	22000	35000 2021
## 1622	22000	35000 2021
## 1623	24000	35000 2021
## 1624	17000	24000 2021
## 1625	21000	35000 2021
## 1626	16000	17000000 2021
## 1627	15000	28000 2021
## 1628	22000	35000 2021
## 1629	130	26000 2021
## 1630	7900	350000 2021
## 1631	21000	28000 2021
## 1632	21000	28000 2021
## 1633	14000	17000 2021
## 1634	21000	35000 2021
## 1635	22000	35000 2021
## 1636	2	1400 2021
## 1637	110	430 2021
## 1638	2	12 2021
## 1639	6	62 2021
## 1640	94	350 2021
## 1641	2	58 2021
## 1642	70	220 2021
## 1643	3	43 2021

```

## 1644          2        24 2021
## 1645         40       280 2021
## 1646         63       280 2021
## 1647         47       220 2021
## 1648         25       210 2021
## 1649         58       350 2021
## 1650         94       240 2021
## 1651        110      240 2021
## 1652        140      350 2021
## 1653        210      430 2021
## 1654        150      280 2021
## 1655        150      280 2021
## 1656         98      240 2021
## 1657        150      280 2021
## 1658        280      350 2021
## 1659        350      540 2021
## 1660        150      430 2021
## 1661        210      350 2021
## 1662        110      280 2021
## 1663        150      280 2021
## 1664         94      240 2021
## 1665        110      240 2021

```

```

View(AirQualityDataset)

AirQualityDataset$State=toupper(AirQualityDataset$State)
AirQualityDataset$City=toupper(AirQualityDataset$City)

AirQualityDataset$NO2=as.numeric(AirQualityDataset$NO2)
AirQualityDataset$PM10=as.numeric(AirQualityDataset$PM10)
AirQualityDataset$S02=as.numeric(AirQualityDataset$S02)
AirQualityDataset$PM2.5=as.numeric(AirQualityDataset$PM2.5)

AirQualityDataset$year=as.integer(AirQualityDataset$year)

finalDataset = merge(WaterQualityData, AirQualityDataset, by = c("State","City","year"), all.x = TRUE)

grouped_df1 = aggregate(cbind(NO2, PM10,PM2.5,S02) ~ State + year, data = AirQualityDataset, mean)
grouped_df2 = aggregate(cbind(NO2, PM10,PM2.5,S02) ~ State, data = AirQualityDataset, mean)

finalDataset = merge(finalDataset, grouped_df1, by = c("State","year"), all.x = TRUE)
finalDataset = merge(finalDataset, grouped_df2, by = c("State"), all.x = TRUE)

fd=finalDataset

finalDataset = subset(fd, select = -c(21, 22, 23, 24, 25, 26, 27, 28))

```

```

finalDataset[finalDataset == "-"] = NA

finalDataset = subset(finalDataset, select = -StateCode)

colSums(is.na(finalDataset))

##          State      year      City TemperatureMin
##          0          0          0            42
## TemperatureMax DissolvedO2Min DissolvedO2Max pHMin
##          37          10          10            0
##          pHMax ConductivityMin ConductivityMax BODMin
##          0          10          10            7
##          BODMax NitrateMin NitrateMax FecalColiformMin
##          7          143          143           109
##          FecalColiformMax TotalColiformMin TotalColiformMax NO2
##          109          49          49            116
##          PM10        PM2.5        S02
##          116          116          116

```

```

waterQuality_2017=finalDataset[finalDataset$year == 2017,]

waterQuality_2017$TemperatureMin      =as.numeric(waterQuality_2017$TemperatureMin)
waterQuality_2017$TemperatureMax     =as.numeric(waterQuality_2017$TemperatureMin)
waterQuality_2017$DissolvedO2Min    =as.numeric(waterQuality_2017$DissolvedO2Min)

## Warning: NAs introduced by coercion

waterQuality_2017$DissolvedO2Max      =as.numeric(waterQuality_2017$DissolvedO2Max)
waterQuality_2017$pHMin                =as.numeric(waterQuality_2017$pHMin)
waterQuality_2017$pHMax                =as.numeric(waterQuality_2017$pHMax)
waterQuality_2017$ConductivityMin     =as.numeric(waterQuality_2017$ConductivityMin)
waterQuality_2017$ConductivityMax     =as.numeric(waterQuality_2017$ConductivityMax)
waterQuality_2017$BODMin               =as.numeric(waterQuality_2017$BODMin)

## Warning: NAs introduced by coercion

waterQuality_2017$BODMax              =as.numeric(waterQuality_2017$BODMax)

## Warning: NAs introduced by coercion

waterQuality_2017$NitrateMin          =as.numeric(waterQuality_2017$NitrateMin)

## Warning: NAs introduced by coercion

waterQuality_2017$NitrateMax          =as.numeric(waterQuality_2017$NitrateMax)
waterQuality_2017$FecalColiformMin    =as.numeric(waterQuality_2017$FecalColiformMin)
waterQuality_2017$FecalColiformMax    =as.numeric(waterQuality_2017$FecalColiformMax)
waterQuality_2017$TotalColiformMin    =as.numeric(waterQuality_2017$TotalColiformMin)

```

```

waterQuality_2017$TotalColiformMax = as.numeric(waterQuality_2017$TotalColiformMax)
waterQuality_2017$NO2 = as.numeric(waterQuality_2017$NO2)
waterQuality_2017$PM10 = as.numeric(waterQuality_2017$PM10)

cols=c("TemperatureMin"
,"TemperatureMax"
,"DissolvedO2Min"
,"DissolvedO2Max"
,"pHMin"
,"pHMax"
,"ConductivityMin"
,"ConductivityMax"
,"BODMin"
,"BODMax"
,"NitrateMin"
,"NitrateMax"
,"FecalColiformMin"
,"FecalColiformMax"
,"TotalColiformMin"
,"TotalColiformMax"
,"NO2"
,"PM10"
,"S02"
,"PM2.5")

waterQuality_2017_Avg=waterQuality_2017

# Calculate the average values for each variable
waterQuality_2017_Avg$TemperatureAvg = (waterQuality_2017_Avg$TemperatureMin + waterQuality_2017_Avg$TemperatureMax) / 2
waterQuality_2017_Avg$DissolvedO2Avg = (waterQuality_2017_Avg$DissolvedO2Min + waterQuality_2017_Avg$DissolvedO2Max) / 2
waterQuality_2017_Avg$pHAvg = (waterQuality_2017_Avg$pHMin + waterQuality_2017_Avg$pHMax) / 2
waterQuality_2017_Avg$ConductivityAvg = (waterQuality_2017_Avg$ConductivityMin + waterQuality_2017_Avg$ConductivityMax) / 2
waterQuality_2017_Avg$FecalColiformAvg = (waterQuality_2017_Avg$FecalColiformMin + waterQuality_2017_Avg$FecalColiformMax) / 2
waterQuality_2017_Avg$BODAvg = (waterQuality_2017_Avg$BODMin + waterQuality_2017_Avg$BODMax) / 2
waterQuality_2017_Avg$NitrateAvg = (waterQuality_2017_Avg$NitrateMin + waterQuality_2017_Avg$NitrateMax) / 2

waterQuality_2017_Avg$TotalColiformAvg = (waterQuality_2017_Avg$TotalColiformMin + waterQuality_2017_Avg$TotalColiformMax) / 2

waterQuality_2017_Avg = subset(waterQuality_2017_Avg,select=-c(TemperatureMin
, TemperatureMax
,DissolvedO2Min
,DissolvedO2Max
,pHMin
,pHMax
,ConductivityMin
,ConductivityMax
,BODMin
,BODMax
,NitrateMin
,NitrateMax
,FecalColiformMin
,FecalColiformMax
))

```

```

,TotalColiformMin
,TotalColiformMax))

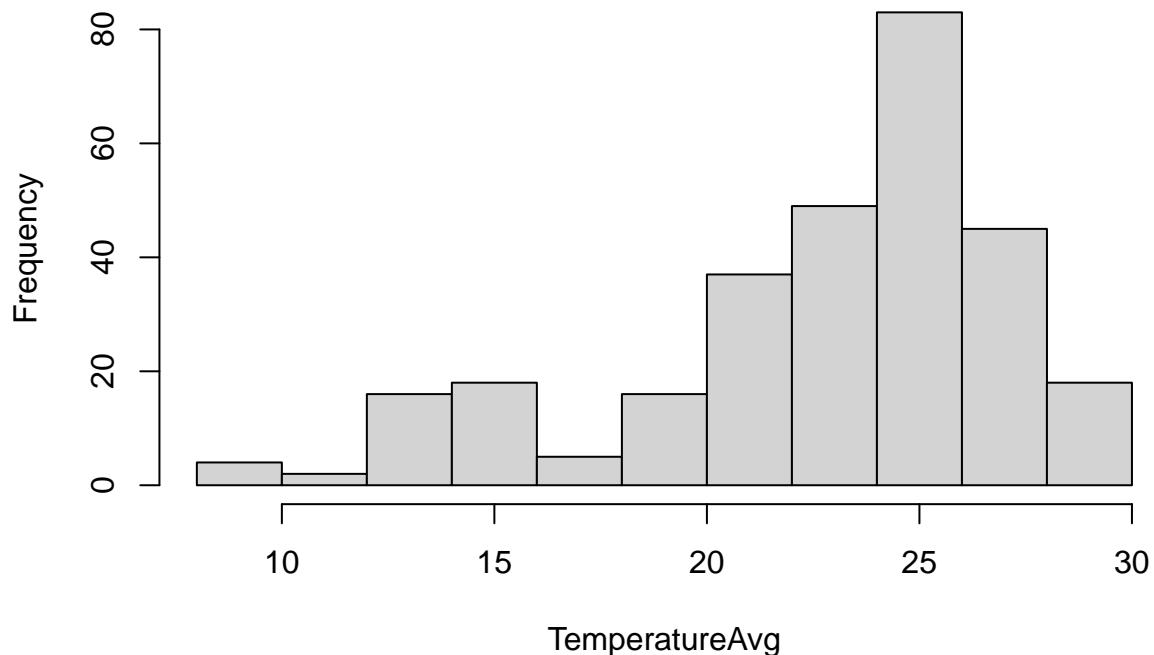
numCols=c("TemperatureAvg"
,"DissolvedO2Avg"
,"pHAvg"
,"ConductivityAvg"
,"BODAvg"
,"NitrateAvg"
,"FecalColiformAvg"
,"TotalColiformAvg"
,"NO2"
,"PM10"
,"SO2"
,"PM2.5")

for (col in colnames(waterQuality_2017_Avg)) {
  # Check if column has any NAs
  if (sum(is.na(waterQuality_2017_Avg[, col])) > 0) {
    # Replace NAs with mean value of the column
    waterQuality_2017_Avg[is.na(waterQuality_2017_Avg[, col]), col] = mean(waterQuality_2017_Avg[, col])
  }
}

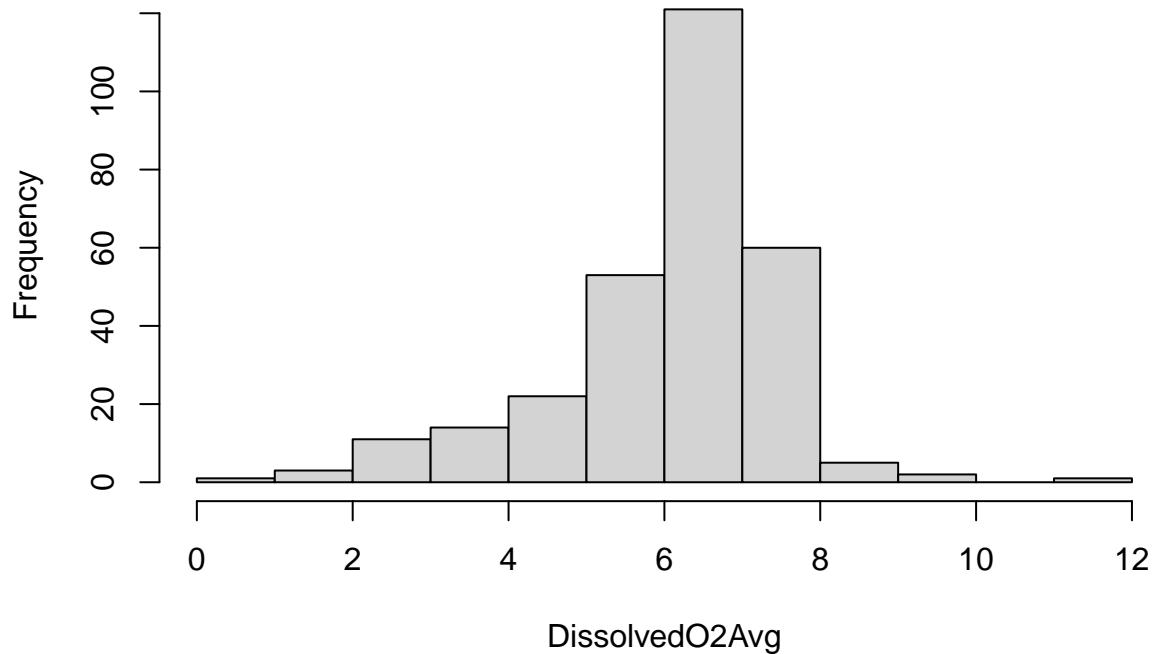
for (col in numCols) {
  # Create the histogram plot
  hist(waterQuality_2017_Avg[[col]], main = col, xlab = col)
}

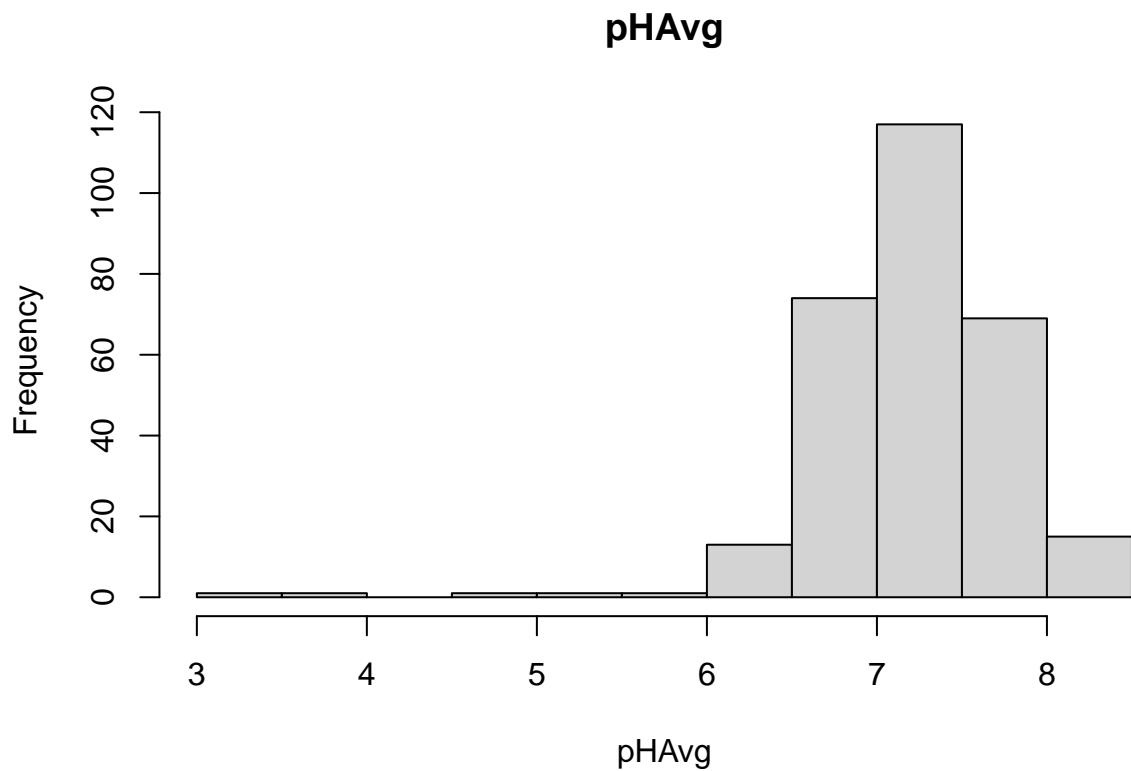
```

TemperatureAvg

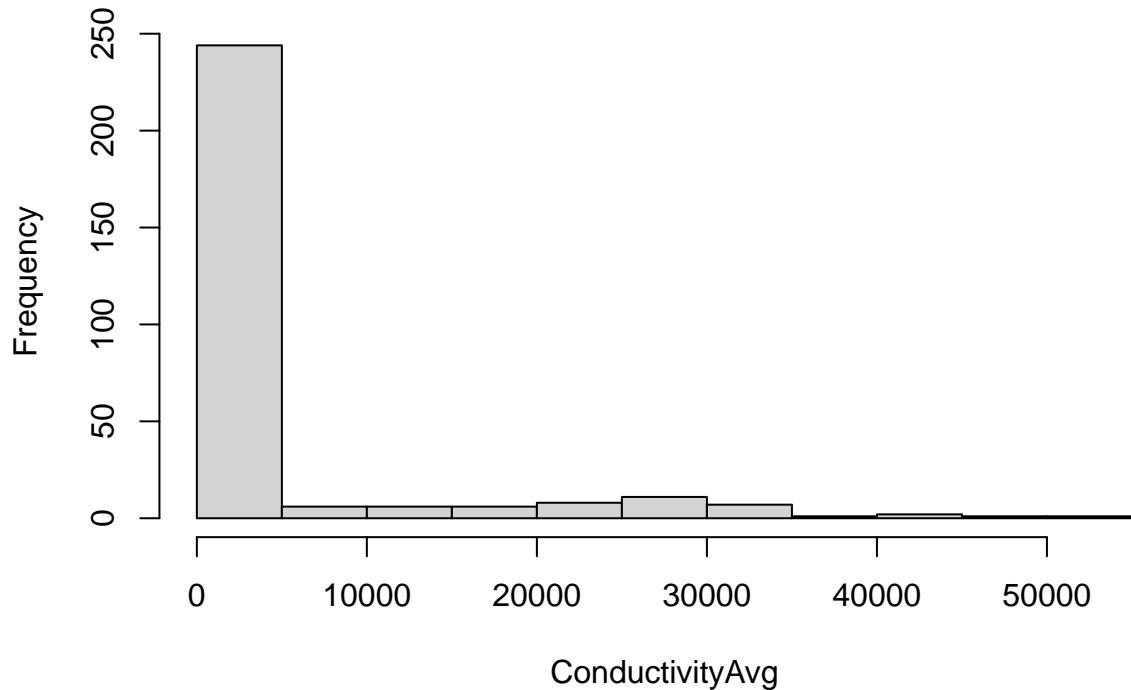


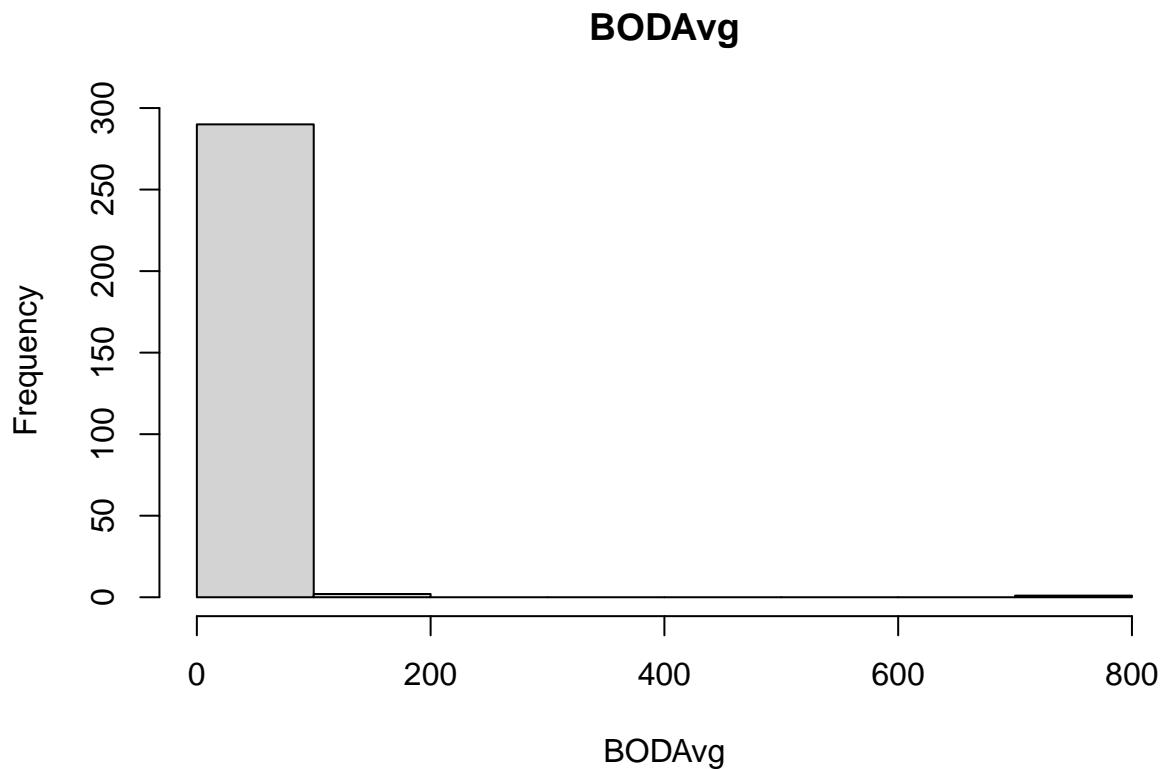
DissolvedO2Avg

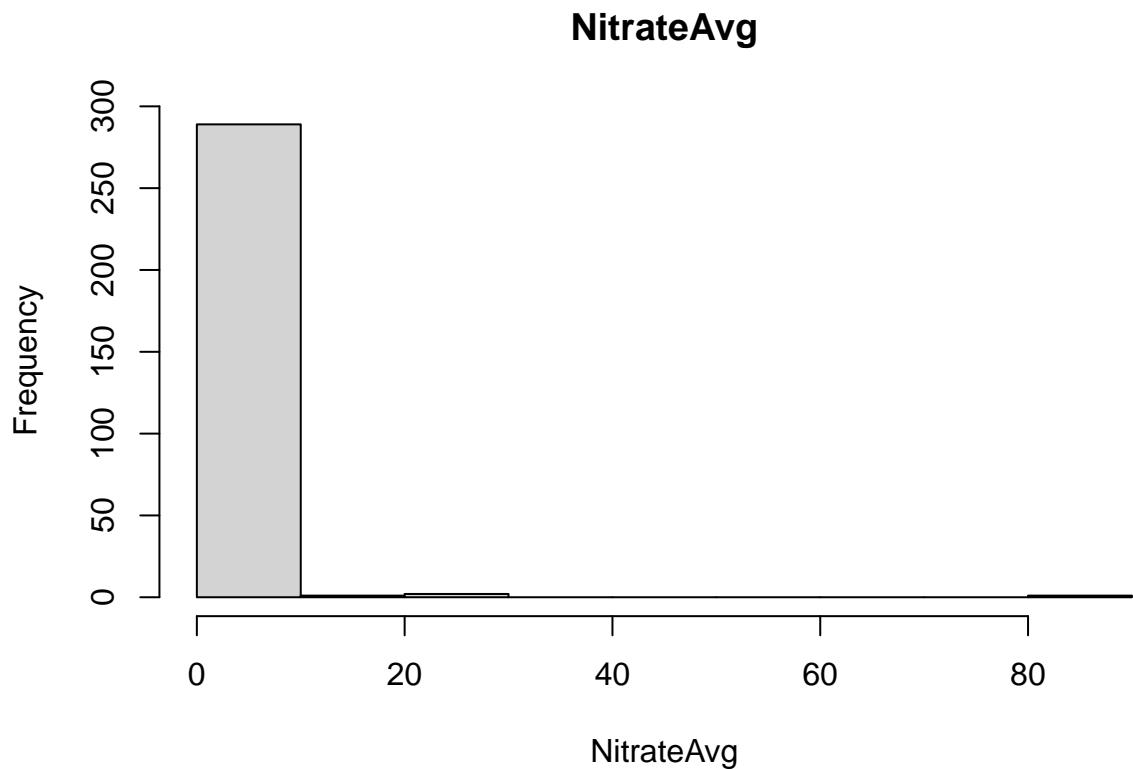




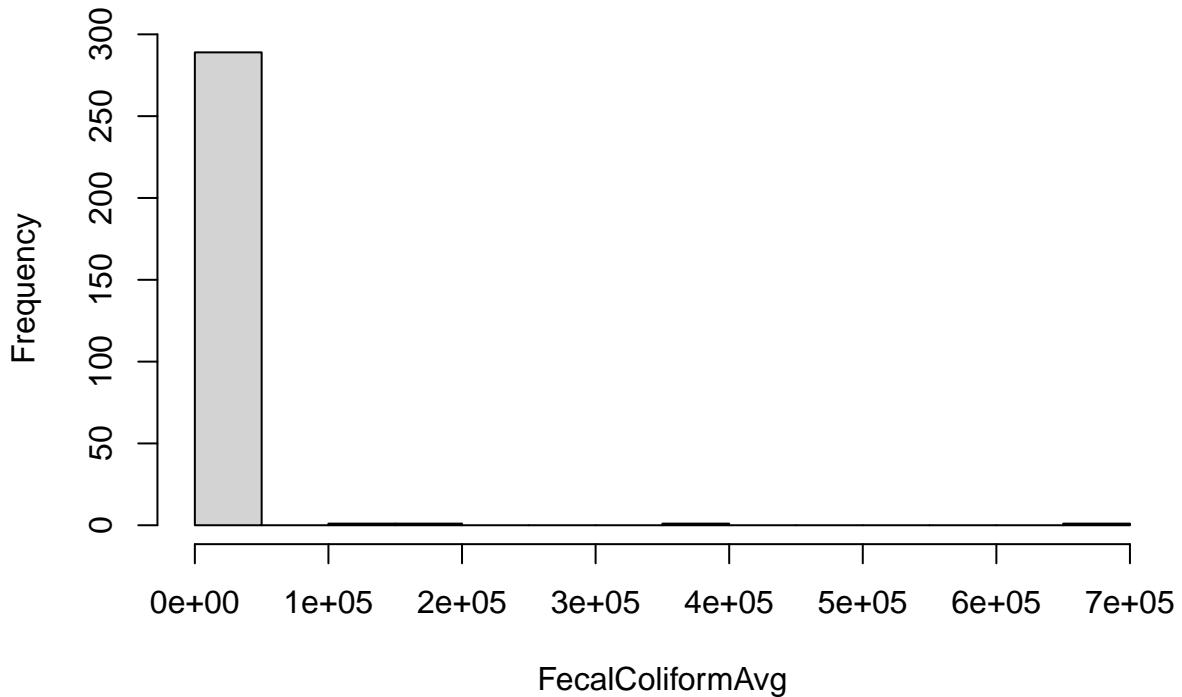
ConductivityAvg

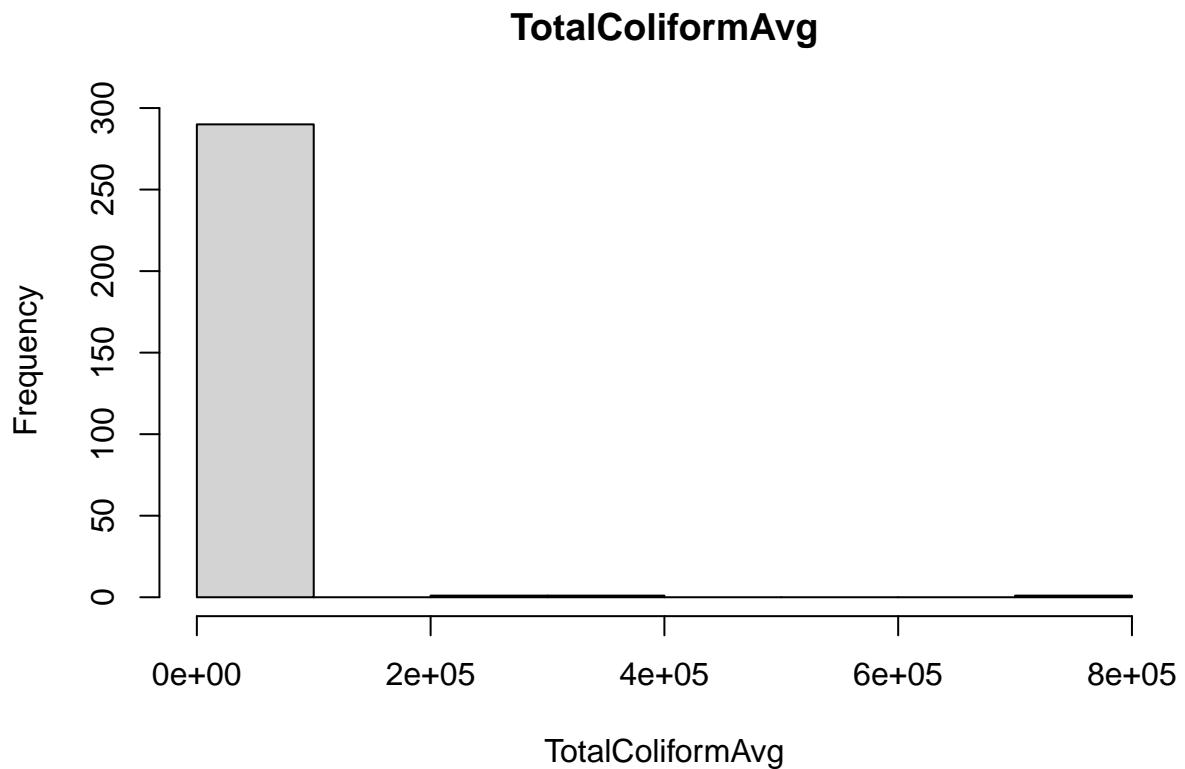




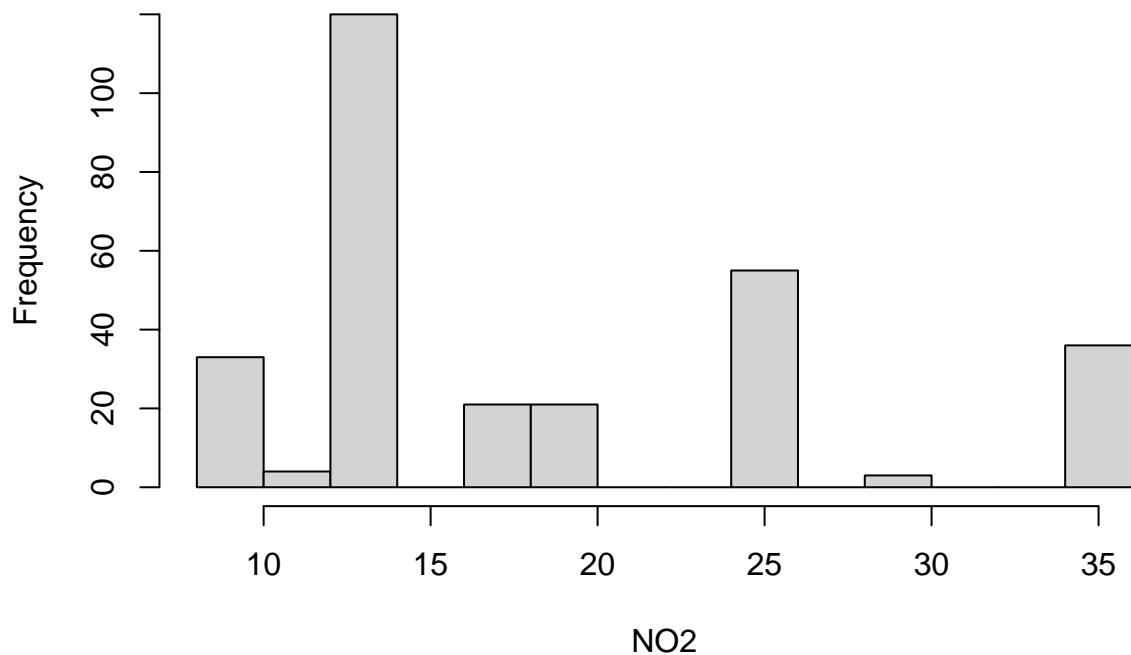


FecalColiformAvg

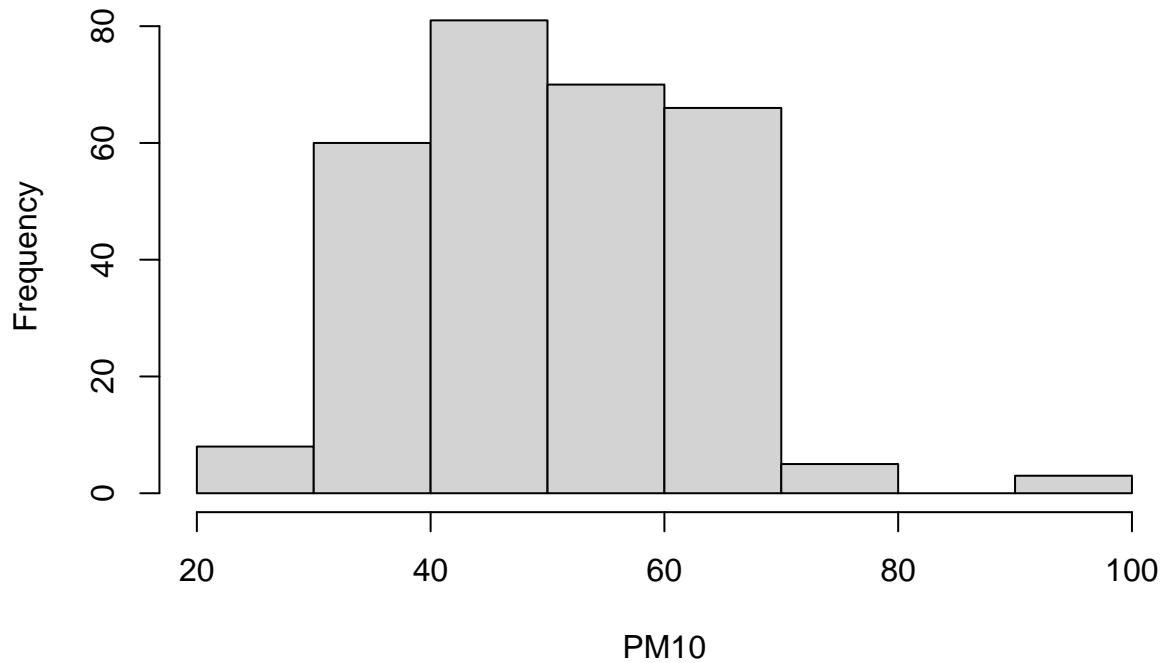




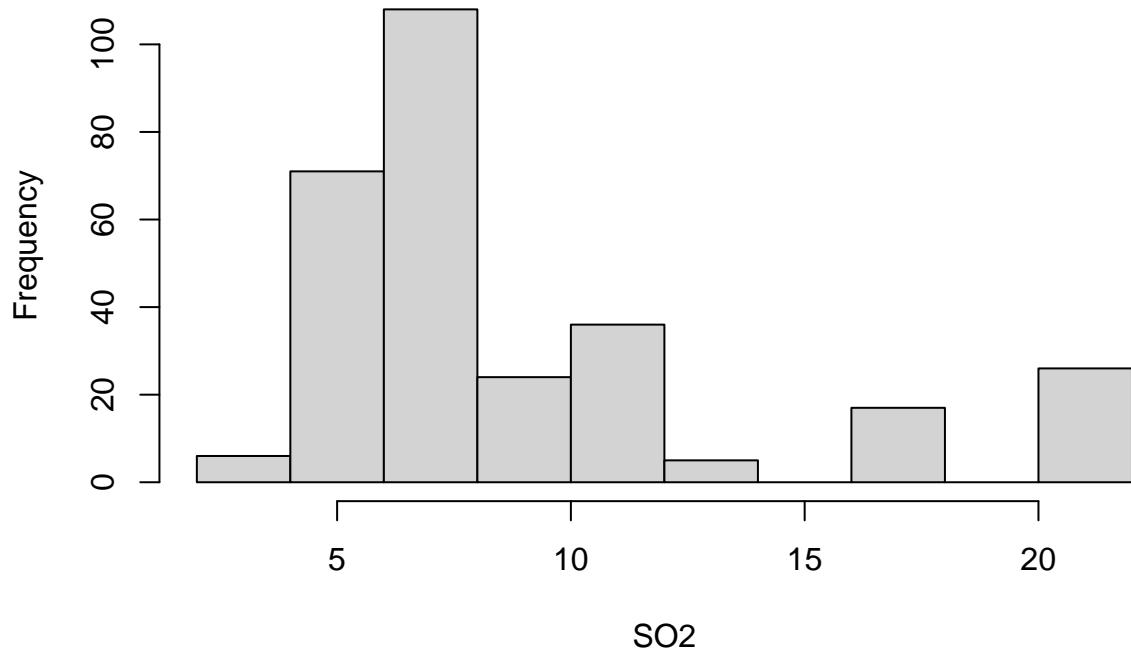
NO2

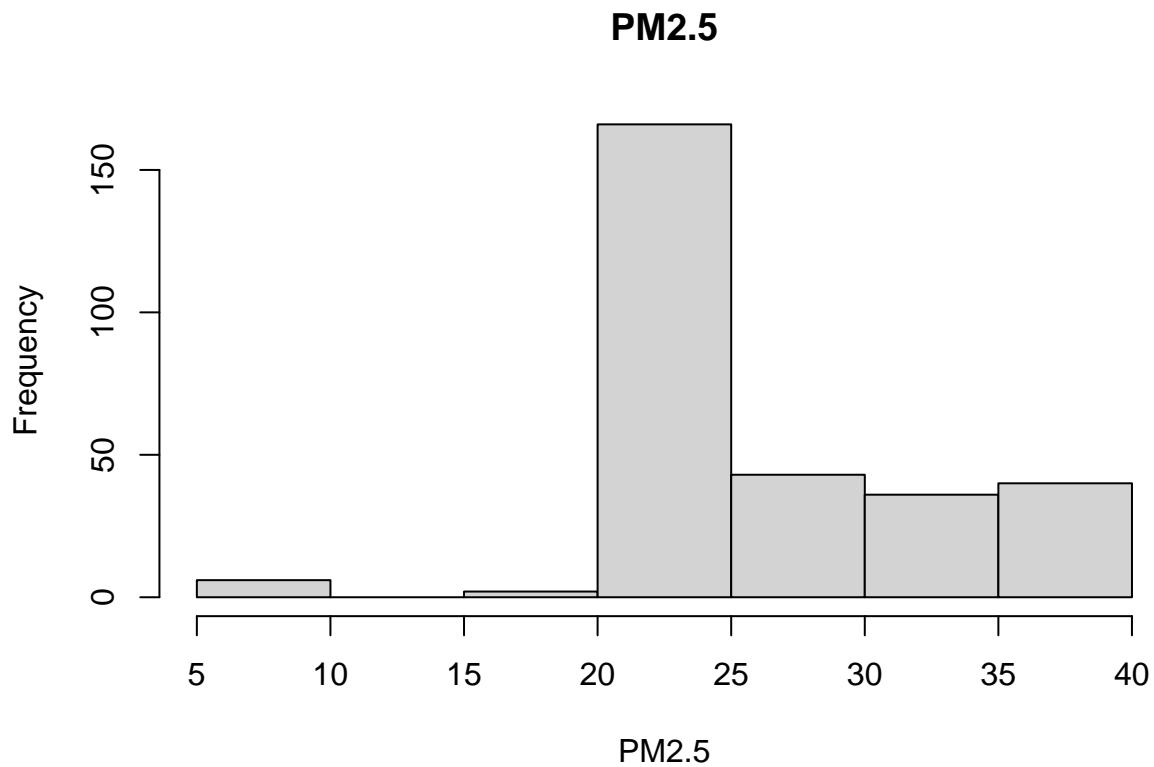


PM10



SO₂





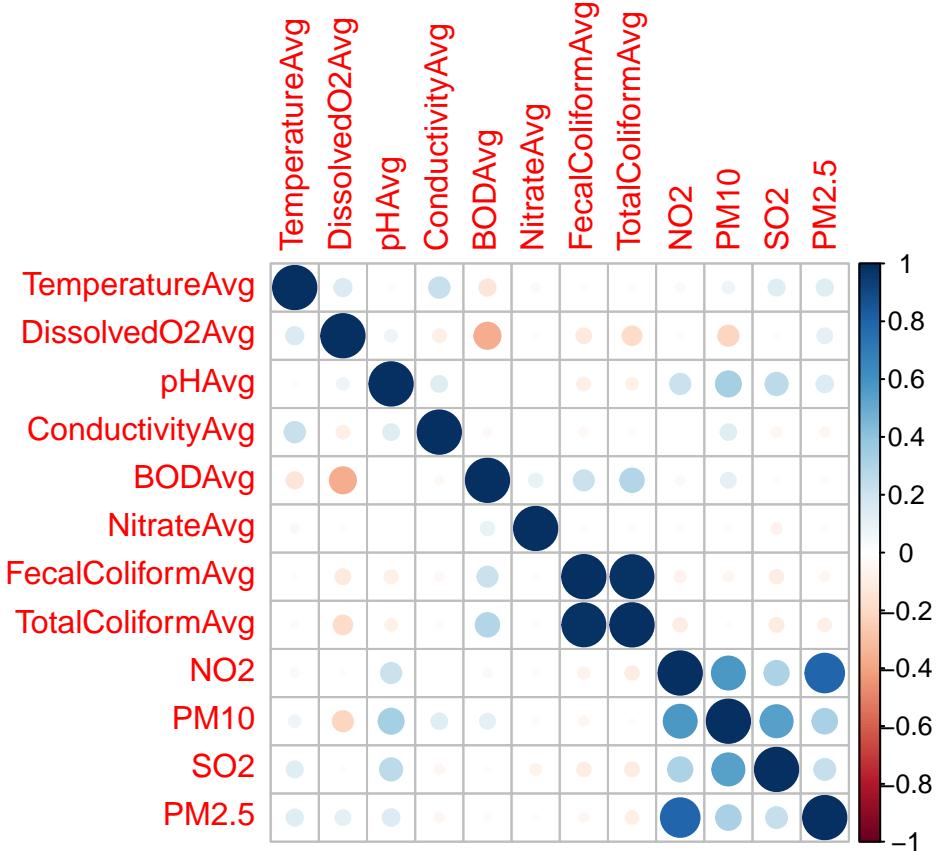
```
cov_mat = cov(waterQuality_2017_Avg[, numCols])
```

```
corr_matrix = cor(waterQuality_2017_Avg[, numCols])
```

```
library(corrplot)
```

```
## corrplot 0.92 loaded
```

```
corrplot(corr_matrix)
```



```
waterQuality_2017_Avg = aggregate(cbind(TemperatureAvg,
                                         DissolvedO2Avg,
                                         pHAvg,
                                         ConductivityAvg,
                                         BODAvg,
                                         NitrateAvg,
                                         FecalColiformAvg,
                                         TotalColiformAvg,
                                         NO2,
                                         PM10,
                                         SO2,
                                         PM2.5) ~ State,
                                     data = waterQuality_2017_Avg,
                                     mean)
```

```
WaterQualityAlone = waterQuality_2017_Avg[, c("TemperatureAvg"
, "DissolvedO2Avg"
, "pHAvg"
, "ConductivityAvg"
, "BODAvg"
, "FecalColiformAvg"
, "NitrateAvg"
, "TotalColiformAvg"
)]
```

```
# Center and scale the data
```

```

WaterQualityAlone_scaled = scale(WaterQualityAlone)

# Perform PCA
pca_result = princomp(WaterQualityAlone_scaled, center = TRUE, scale. = TRUE)

## Warning: In princomp.default(WaterQualityAlone_scaled, center = TRUE, scale. = TRUE) :
##   extra arguments 'center', 'scale.' will be disregarded

# Print the summary of PCA results
summary(pca_result)

## Importance of components:
##                               Comp.1    Comp.2    Comp.3    Comp.4    Comp.5
## Standard deviation      1.8146658 1.1892782 1.1439859 0.85904559 0.57131965
## Proportion of Variance 0.4390683 0.1885843 0.1744938 0.09839458 0.04352082
## Cumulative Proportion  0.4390683 0.6276526 0.8021465 0.90054104 0.94406186
##                               Comp.6    Comp.7    Comp.8
## Standard deviation      0.50050392 0.3601625 0.198279855
## Proportion of Variance 0.03340056 0.0172956 0.005241987
## Cumulative Proportion  0.97746241 0.9947580 1.000000000

#reverse the signs
pca_result$rotation = -1*pca_result$rotation

#display principal components
pca_result$rotation

## numeric(0)

var_explained = pca_result$sdev^2 / sum(pca_result$sdev^2)

print(var_explained)

##           Comp.1    Comp.2    Comp.3    Comp.4    Comp.5    Comp.6
## 0.439068273 0.188584344 0.174493843 0.098394577 0.043520818 0.033400556
##           Comp.7    Comp.8
## 0.017295602 0.005241987

pc.comp = pca_result$scores

#87%
pc.comp1 = -1*pc.comp[,1]
pc.comp2 = -1*pc.comp[,2]
pc.comp3 = -1*pc.comp[,3]
pc.comp4 = -1*pc.comp[,4]
pc.comp5 = -1*pc.comp[,5]
pc.comp6 = -1*pc.comp[,6]

library(factoextra)

## Loading required package: ggplot2

```

```

## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa

library(tidyverse)
library(ggplot2)

X = cbind(pc.comp1, pc.comp2, pc.comp3, pc.comp4, pc.comp5, pc.comp6)

X

##          pc.comp1    pc.comp2    pc.comp3    pc.comp4    pc.comp5    pc.comp6
## [1,] -1.0094213  0.53611506 -1.3197288  0.03375428  0.19437300 -1.09453633
## [2,] -1.3967625 -1.44339630  0.7873702  1.80466323  0.13753972 -0.41329506
## [3,]  0.2119474 -2.46856150 -1.4411109  0.82752166 -0.32684075  0.55999324
## [4,] -0.8599363 -0.55025993  1.3263095  0.29844103 -1.32036263  0.38128402
## [5,]  2.1681169  0.01821827  0.4830776 -0.59348019 -0.85976354 -0.70485271
## [6,] -0.3673538  1.72746067  0.3008697 -0.02546562 -0.10007942  0.95061585
## [7,] -1.3562605  0.38691127 -0.3104345 -0.92898550 -0.15596234 -0.47337758
## [8,]  1.3112576 -1.32662550 -1.8332806 -0.43162172  0.30978745  0.19410837
## [9,] -0.8866027  0.22844910 -0.1826866  0.34741078 -0.25500835 -0.31679893
## [10,] -0.1162755  0.65408640 -1.1798106 -0.49762819  0.10355544  0.15743170
## [11,]  1.2231685  1.28116053 -1.7366682  0.36100005  0.31473338  0.45105874
## [12,] -1.5791522 -1.59707793  1.5659784 -1.00096738  1.32932737  0.10180988
## [13,]  5.5614610 -0.25919816  1.4968510 -0.26850722  0.28659847 -0.08506035
## [14,]  0.1591854  1.94111648  0.9180136  1.81301040  0.60059920 -0.07377380
## [15,] -1.7582087  0.23377406  0.9812632 -0.84348270 -0.01108269  0.40588091
## [16,] -1.3051633  0.63782748  0.1439869 -0.89566293 -0.24741432 -0.04048797

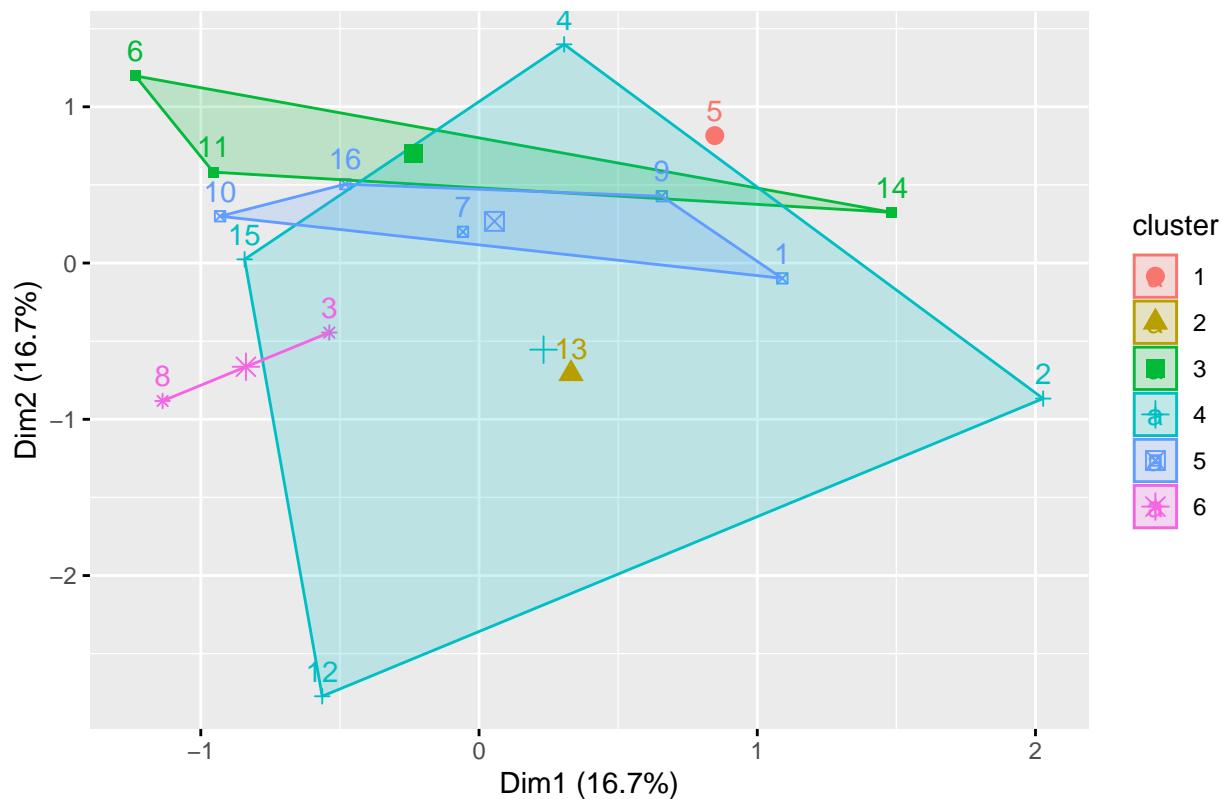
k = 6
kmeans_model = kmeans(X, centers = k, nstart=1)

waterQuality_2017_Avg$cluster=kmeans_model$cluster

fviz_cluster(kmeans_model, data = X)

```

Cluster plot



```
mean_by_group = aggregate(cbind(TemperatureAvg, DissolvedO2Avg, pHAvg, ConductivityAvg, BODAvg, NitrateAvg),
                          by = list(waterQuality_2017_Avg$State), FUN = mean)
table(waterQuality_2017_Avg$State, waterQuality_2017_Avg$cluster)
```

```
##                                     1 2 3 4 5 6
## ANDHRA PRADESH                  0 0 0 0 1 0
## DAMAN AND DIU, DADRA\r\nAND NAGAR HAVELI 0 0 0 1 0 0
## GOA                           0 0 0 0 0 1
## GUJARAT                      0 0 0 1 0 0
## HARYANA                       1 0 0 0 0 0
## HIMACHAL PRADESH                0 0 1 0 0 0
## KARNATAKA                     0 0 0 0 1 0
## KERALA                         0 0 0 0 0 1
## MAHARASHTRA                    0 0 0 0 1 0
## MANIPUR                        0 0 0 0 1 0
## MEGHALAYA                      0 0 1 0 0 0
## PUDUCHERRY                     0 0 0 1 0 0
## PUNJAB                         0 1 0 0 0 0
## RAJASTHAN                      0 0 1 0 0 0
## TAMIL NADU                     0 0 0 1 0 0
## TRIPURA                        0 0 0 0 1 0
```

```
WaterQualityAlone
```

```

##      TemperatureAvg DissolvedO2Avg      pHAvg ConductivityAvg      BODAvg
## 1      26.00000    7.100000 7.200000      462.5000 1.775000
## 2      26.11111    6.816667 8.005556     13533.6667 6.144444
## 3      26.75217    6.458696 7.071739     15039.6739 2.486957
## 4      24.58824    5.114706 7.644118      9336.4118 11.038235
## 5      22.98681    4.440000 7.320000     1469.9000 49.920000
## 6      14.00000    6.441667 7.350000      843.8333 14.633333
## 7      27.00000    6.808333 7.283333      137.5000 2.758333
## 8      25.71690    5.888732 6.950000     4915.1761 2.775352
## 9      23.50000    6.504492 7.416667     4500.4583 8.186111
## 10     21.25758    6.572727 7.050000      328.5303 7.139394
## 11     14.51852    6.214815 6.907407      198.8148 9.918519
## 12     29.50000    7.900000 8.075000      5908.5000 9.338235
## 13     14.92308    3.208143 7.688462      887.0769 75.826923
## 14     13.00000    5.216667 8.033333      566.6667 1.400000
## 15     24.41667    6.862500 7.629167     2651.2500 2.262500
## 16     24.78846    6.461538 7.369231      135.3462 1.109615
##      FecalColiformAvg NitrateAvg TotalColiformAvg
## 1      18.500000  1.7725000 342.50000
## 2      480.888889 1.4544444 785.33333
## 3      10291.000000 1.6443968 14351.82609
## 4      149.794118 0.4597059 812.94118
## 5      7606.787747 1.3161265 14633.61503
## 6      42.250000 0.7758333 108.58333
## 7      269.833333 0.8466667 717.41667
## 8      15450.626761 1.6994366 17967.12676
## 9      69.041667 1.2793368 165.95833
## 10     6909.646245 1.3161265 410.19697
## 11     5736.814815 1.9396296 12302.44444
## 12     6909.646245 0.1925000 8974.35839
## 13     17319.230769 1.6226923 40500.00000
## 14     8.666667 1.6866667 25.33333
## 15     66.958333 0.1800000 124.29167
## 16     192.500000 0.5936538 322.69231

```

```
Result_2017_water<- mean_by_group[order(mean_by_group$TotalColiformAvg, mean_by_group$BODAvg, mean_by_group$FecalColiformAvg),]
```

```
Result_2017_water$Rank=c("A", "B", "C", "D", "E", "F")
```

```
Water2017 = merge(waterQuality_2017_Avg, Result_2017_water[,c("cluster","Rank")], by = "cluster", all.x=TRUE)
```

```
pca_data = waterQuality_2017_Avg[, c("TemperatureAvg",
,"DissolvedO2Avg",
,"pHAvg",
,"ConductivityAvg",
,"BODAvg",
,"FecalColiformAvg",
,"NitrateAvg",
,"TotalColiformAvg",
,"NO2"]]
```

```

,"PM10"
,"SO2"
,"PM2.5")]

# Center and scale the data
pca_data_scaled = scale(pca_data)

# Perform PCA
pca_result = princomp(pca_data_scaled, center = TRUE, scale. = TRUE)

## Warning: In princomp.default(pca_data_scaled, center = TRUE, scale. = TRUE) :
##   extra arguments 'center', 'scale.' will be disregarded

# Print the summary of PCA results
summary(pca_result)

## Importance of components:
##                               Comp.1    Comp.2    Comp.3    Comp.4    Comp.5
## Standard deviation      1.8552403 1.6513287 1.2364663 1.03617229 0.91758906
## Proportion of Variance 0.3059481 0.2423899 0.1358977 0.09543582 0.07484175
## Cumulative Proportion  0.3059481 0.5483380 0.6842357 0.77967154 0.85451329
##                               Comp.6    Comp.7    Comp.8    Comp.9    Comp.10
## Standard deviation     0.8732970 0.6062998 0.48530935 0.39403716 0.246568352
## Proportion of Variance 0.0677909 0.0326755 0.02093557 0.01380136 0.005404085
## Cumulative Proportion  0.9223042 0.9549797 0.97591526 0.98971662 0.995120706
##                               Comp.11   Comp.12
## Standard deviation     0.175338638 0.155397616
## Proportion of Variance 0.002732768 0.002146526
## Cumulative Proportion  0.997853474 1.000000000

#reverse the signs
pca_result$rotation = -1*pca_result$rotation

#display principal components
pca_result$rotation

## numeric(0)

var_explained = pca_result$sdev^2 / sum(pca_result$sdev^2)

print(var_explained)

##      Comp.1    Comp.2    Comp.3    Comp.4    Comp.5    Comp.6
## 0.305948142 0.242389907 0.135897671 0.095435823 0.074841750 0.067790896
##      Comp.7    Comp.8    Comp.9    Comp.10   Comp.11   Comp.12
## 0.032675504 0.020935570 0.013801358 0.005404085 0.002732768 0.002146526

pc.comp = pca_result$scores

#84%

```

```

pc.comp1 = -1*pc.comp[,1]
pc.comp2 = -1*pc.comp[,2]
pc.comp3 = -1*pc.comp[,3]
pc.comp4 = -1*pc.comp[,4]
pc.comp5= -1*pc.comp[,5]

library(factoextra)
library(tidyverse)
library(ggplot2)

X = cbind(pc.comp1, pc.comp2,pc.comp3, pc.comp4,pc.comp5)

X

##          pc.comp1    pc.comp2    pc.comp3    pc.comp4    pc.comp5
## [1,] -1.0257589 -1.88775530  0.8945993 -0.7289317 -0.21517961
## [2,] -1.3735625  0.49327406 -1.2983128 -0.3173637  1.32496235
## [3,]  0.3541202 -1.25390009 -2.0651711 -1.3932519 -0.02341779
## [4,] -0.9429651  2.24432940 -0.3488594  0.3798702 -0.71510228
## [5,]  2.2114628  1.59151657  0.1276494  0.0633817 -1.07439800
## [6,] -0.1115279 -2.81421913  1.9585091  1.5311612  1.00785382
## [7,] -1.3395268 -0.64183608  0.4082663  0.2172671 -0.43592197
## [8,]  1.2842255 -1.93655785 -1.3502824 -1.0665042 -0.44572403
## [9,] -0.9524811  2.13950433  0.3410365 -1.4174892 -0.06782937
## [10,] -0.2978967 -0.43485477  0.4723653 -1.0005260 -0.22597948
## [11,]  1.1962888 -2.06530411  1.1393566 -0.6626610  0.08179564
## [12,] -1.8012698  0.08999583 -2.4315783  1.6873941  0.91426213
## [13,]  5.6054007  0.64569241 -0.5430013  1.3281224  0.14374165
## [14,]  0.4856780  2.94672095  1.8246916 -0.7558295  1.78884303
## [15,] -1.7541699  0.25653933  0.1243600  1.1571537  0.05792929
## [16,] -1.5380176  0.62685444  0.7463713  0.9782068 -2.11583538

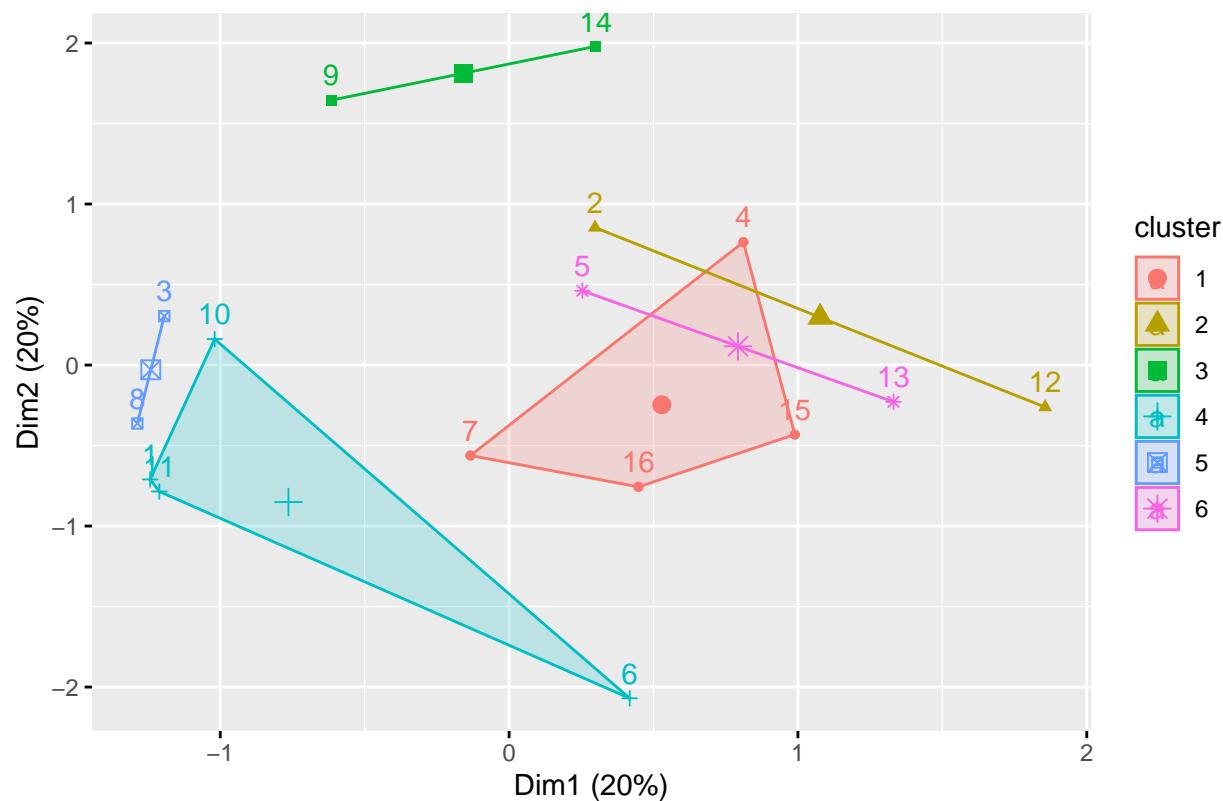
k =6
kmeans_model = kmeans(X, centers = k,nstart=1)

waterQuality_2017_Avg$cluster=kmeans_model$cluster

fviz_cluster(kmeans_model, data = X)

```

Cluster plot



```
mean_by_group = aggregate(cbind(TemperatureAvg, DissolvedO2Avg, pHAvg, ConductivityAvg, BODAvg, NitrateAvg),
                          by = list(waterQuality_2017_Avg$State), FUN = mean)
table(waterQuality_2017_Avg$State, waterQuality_2017_Avg$cluster)
```

```
##
##                                     1 2 3 4 5 6
## ANDHRA PRADESH                  0 0 0 1 0 0
## DAMAN AND DIU, DADRA\r\nAND NAGAR HAVELI 0 1 0 0 0 0
## GOA                               0 0 0 0 1 0
## GUJARAT                           1 0 0 0 0 0
## HARYANA                            0 0 0 0 0 1
## HIMACHAL PRADESH                 0 0 0 1 0 0
## KARNATAKA                         1 0 0 0 0 0
## KERALA                            0 0 0 0 1 0
## MAHARASHTRA                        0 0 1 0 0 0
## MANIPUR                            0 0 0 1 0 0
## MEGHALAYA                          0 0 0 1 0 0
## PUDUCHERRY                         0 1 0 0 0 0
## PUNJAB                             0 0 0 0 0 1
## RAJASTHAN                           0 0 1 0 0 0
## TAMIL NADU                          1 0 0 0 0 0
## TRIPURA                            1 0 0 0 0 0
```

```
waterQuality_2017_Avg
```

```

##                                     State TemperatureAvg DissolvedO2Avg
## 1          ANDHRA PRADESH           26.00000    7.100000
## 2 DAMAN AND DIU, DADRA\r\nAND NAGAR HAVELI 26.11111    6.816667
## 3                               GOA           26.75217    6.458696
## 4                               GUJARAT        24.58824    5.114706
## 5                               HARYANA        22.98681    4.440000
## 6 HIMACHAL PRADESH           14.00000    6.441667
## 7          KARNATAKA           27.00000    6.808333
## 8          KERALA             25.71690    5.888732
## 9 MAHARASHTRA                23.50000    6.504492
## 10          MANIPUR            21.25758    6.572727
## 11          MEGHALAYA          14.51852    6.214815
## 12          PUDUCHERRY         29.50000    7.900000
## 13          PUNJAB              14.92308    3.208143
## 14          RAJASTHAN           13.00000    5.216667
## 15          TAMIL NADU          24.41667    6.862500
## 16          TRIPURA              24.78846    6.461538
##      pHAvg ConductivityAvg     BODAvg NitrateAvg FecalColiformAvg
## 1 7.200000       462.5000  1.775000  1.7725000    18.500000
## 2 8.005556       13533.6667 6.144444  1.4544444   480.888889
## 3 7.071739       15039.6739 2.486957  1.6443968  10291.000000
## 4 7.644118       9336.4118 11.038235  0.4597059   149.794118
## 5 7.320000       1469.9000 49.920000  1.3161265   7606.787747
## 6 7.350000       843.8333 14.633333  0.7758333   42.250000
## 7 7.283333       137.5000 2.758333  0.8466667   269.833333
## 8 6.950000       4915.1761 2.775352  1.6994366  15450.626761
## 9 7.416667       4500.4583 8.186111  1.2793368   69.041667
## 10 7.050000      328.5303 7.139394  1.3161265   6909.646245
## 11 6.907407      198.8148 9.918519  1.9396296   5736.814815
## 12 8.075000      5908.5000 9.338235  0.1925000   6909.646245
## 13 7.688462      887.0769 75.826923  1.6226923  17319.230769
## 14 8.033333      566.6667 1.400000  1.6866667   8.666667
## 15 7.629167      2651.2500 2.262500  0.1800000   66.958333
## 16 7.369231      135.3462 1.109615  0.5936538   192.500000
##      TotalColiformAvg      NO2      PM10      S02      PM2.5 cluster
## 1          342.50000 17.562500 28.25000  7.125000 17.562500    4
## 2          785.33333 18.269553 51.77669  9.068159 26.843819    2
## 3          14351.82609 13.411765 59.06424  7.021524 21.260108    5
## 4          812.94118 24.562963 67.99259 16.120823 29.083951    1
## 5          14633.61503 24.576923 78.03846 12.346154 26.329325    6
## 6          108.58333 8.454545 28.45455  2.272727 8.454545    4
## 7          717.41667 17.102624 47.58049  7.824026 24.676111    1
## 8          17967.12676 12.891386 43.55600  4.900582 24.997756    5
## 9          165.95833 34.694938 67.94619 11.944126 35.157701    3
## 10         410.19697 24.231229 38.50000  7.254072 32.640613    4
## 11         12302.44444 9.928571 39.37500  6.562912 23.854899    4
## 12         8974.35839 11.500000 41.83333  6.500000 39.281226    2
## 13        40500.00000 17.832208 67.48132  7.546342 23.521703    6
## 14         25.33333 28.032680 99.91756  9.937908 32.468740    3
## 15         124.29167 18.269553 51.77669  9.068159 26.843819    1
## 16         322.69231 13.000000 55.75000 21.000000 24.250000    1

```

```
Result_2017= mean_by_group[order(mean_by_group$TotalColiformAvg, mean_by_group$BODAvg, mean_by_group$Di
```

```

Result_2017$Rank=c("A", "B", "C", "D", "E","F")
class(Result_2017$cluster)

## [1] "integer"

class(waterQuality_2017_Avg$cluster)

## [1] "integer"

WaterAir2017 = merge(waterQuality_2017_Avg, Result_2017[,c("cluster","Rank")], by = "cluster", all.x = T)

library(factoextra)
library(tidyr)
library(ggplot2)

waterQuality_2017_Avg_clust = waterQuality_2017_Avg[, c("TemperatureAvg"
,"DissolvedO2Avg"
,"pHAvg"
,"ConductivityAvg"
,"BODAvg"
,"FecalColiformAvg"
,"NitrateAvg"
,"TotalColiformAvg"
,"NO2"
,"PM10"
,"S02"
,"PM2.5")]

selected_cols = c("TemperatureAvg"
,"DissolvedO2Avg"
,"pHAvg"
,"ConductivityAvg"
,"BODAvg"
,"FecalColiformAvg"
,"NitrateAvg"
,"TotalColiformAvg"
,"NO2"
,"PM10"
,"S02"
,"PM2.5")

data_selected = waterQuality_2017_Avg[, selected_cols]

scaled_df = waterQuality_2017_Avg
scaled_df = subset(waterQuality_2017_Avg, select = c(State, TemperatureAvg
,DissolvedO2Avg
,pHAvg
,ConductivityAvg
,BODAvg
,FecalColiformAvg

```

```

,NitrateAvg
,TotalColiformAvg
,NO2
,PM10
,S02
,PM2.5))

rownames(scaled_df)=scaled_df$State

k =6
kmeans_model = kmeans(scaled_df[,-1], centers = k,nstart=1)

# Print cluster centers
print(kmeans_model$centers)

##   TemperatureAvg DissolvedO2Avg    pHAvg ConductivityAvg    BODAvg
## 1      26.23454     6.173714 7.010870      9977.4250  2.631154
## 2      22.33511     6.184938 7.434136     2525.7383 23.058918
## 3      21.81502     6.485028 7.468819     1328.2221  4.589270
## 4      21.25758     6.572727 7.050000      328.5303  7.139394
## 5      14.92308     3.208143 7.688462      887.0769 75.826923
## 6      25.34967     5.965686 7.824837     11435.0392  8.591340
##   FecalColiformAvg NitrateAvg TotalColiformAvg      NO2      PM10      S02
## 1      12870.81338  1.6719167     16159.4764 13.15158 51.31012  5.961053
## 2       6751.08294  1.1494187     11970.1393 15.33516 53.08226  8.469689
## 3       95.39286   1.0192368      258.1108 19.58812 54.23935  9.881707
## 4      6909.64625  1.3161265      410.1970 24.23123 38.50000  7.254072
## 5      17319.23077  1.6226923     40500.0000 17.83221 67.48132  7.546342
## 6      315.34150   0.9570752      799.1373 21.41626 59.88464 12.594491
##   PM2.5
## 1 23.12893
## 2 29.82182
## 3 24.20192
## 4 32.64061
## 5 23.52170
## 6 27.96389

# Print cluster membership for each observation
waterQuality_2017_Avg$cluster=kmeans_model$cluster

# create a new column to store the classification

# view the updated dataset with the new classification column
head(waterQuality_2017_Avg)

##                                     State TemperatureAvg DissolvedO2Avg
## 1 ANDHRA PRADESH            26.00000    7.100000
## 2 DAMAN AND DIU, DADRA\r\nAND NAGAR HAVELI 26.11111    6.816667
## 3                               GOA            26.75217    6.458696
## 4                               GUJARAT        24.58824    5.114706

```

```

## 5 HARYANA 22.98681 4.440000
## 6 HIMACHAL PRADESH 14.00000 6.441667
## pHAvg ConductivityAvg BODAvg NitrateAvg FecalColiformAvg
## 1 7.200000 462.5000 1.775000 1.7725000 18.5000
## 2 8.005556 13533.6667 6.144444 1.4544444 480.8889
## 3 7.071739 15039.6739 2.486957 1.6443968 10291.0000
## 4 7.644118 9336.4118 11.038235 0.4597059 149.7941
## 5 7.320000 1469.9000 49.920000 1.3161265 7606.7877
## 6 7.350000 843.8333 14.633333 0.7758333 42.2500
## TotalColiformAvg NO2 PM10 S02 PM2.5 cluster
## 1 342.5000 17.562500 28.25000 7.125000 17.562500 3
## 2 785.3333 18.269553 51.77669 9.068159 26.843819 6
## 3 14351.8261 13.411765 59.06424 7.021524 21.260108 1
## 4 812.9412 24.562963 67.99259 16.120823 29.083951 6
## 5 14633.6150 24.576923 78.03846 12.346154 26.329325 2
## 6 108.5833 8.454545 28.45455 2.272727 8.454545 3

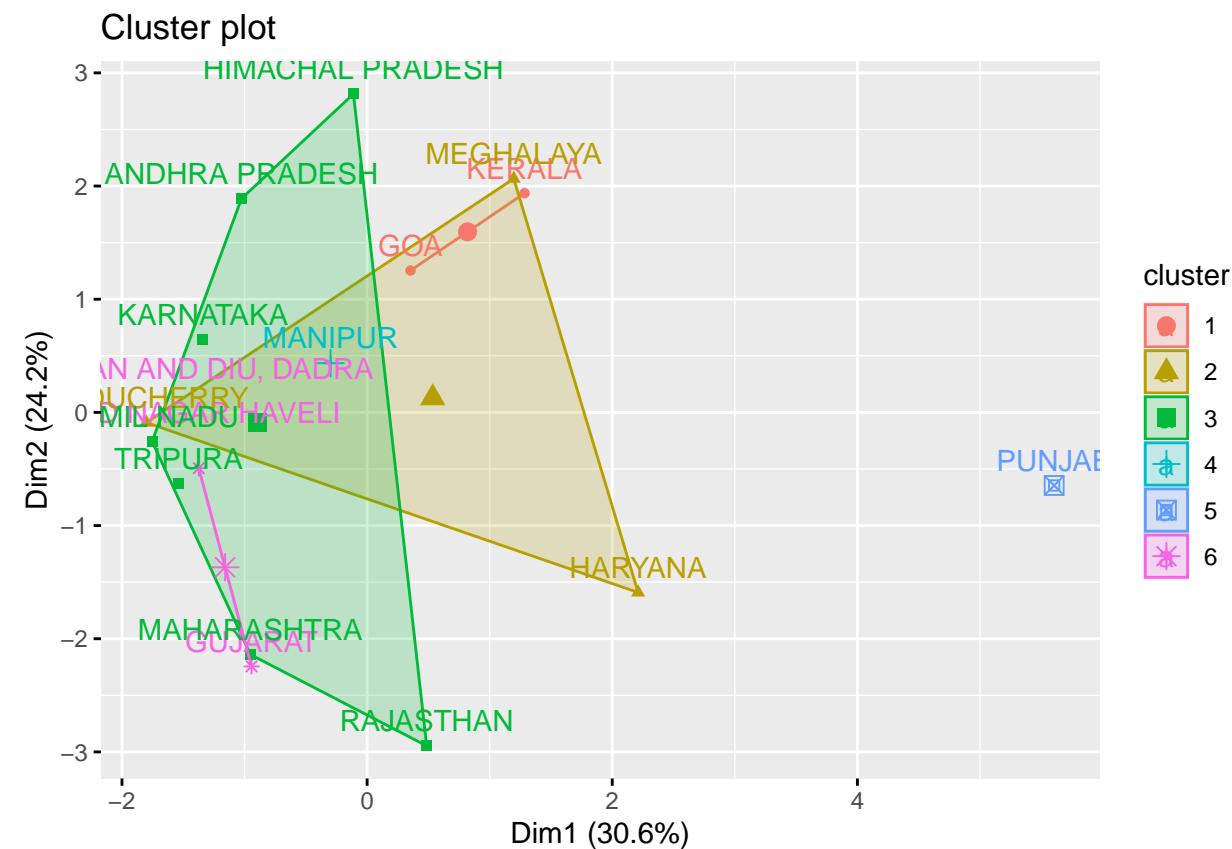
```

```
fviz_cluster(kmeans_model, data = scaled_df[,-1])
```

```

## Warning in grid.Call.graphics(C_text, as.graphicsAnnot(x$label), x$x, x$y, :
## font width unknown for character 0xd

```



```

mean_by_group = aggregate(cbind(TemperatureAvg, DissolvedO2Avg, pHAvg, ConductivityAvg, BODAvg, FecalColif)
table(waterQuality_2017_Avg$State, waterQuality_2017_Avg$cluster)

```

```

##                                     1 2 3 4 5 6
## ANDHRA PRADESH                      0 0 1 0 0 0
## DAMAN AND DIU, DADRA\r\nAND NAGAR HAVELI 0 0 0 0 0 1
## GOA                                    1 0 0 0 0 0
## GUJARAT                                0 0 0 0 0 1
## HARYANA                                 0 1 0 0 0 0
## HIMACHAL PRADESH                      0 0 1 0 0 0
## KARNATAKA                               0 0 1 0 0 0
## KERALA                                   1 0 0 0 0 0
## MAHARASHTRA                            0 0 1 0 0 0
## MANIPUR                                  0 0 0 1 0 0
## MEGHALAYA                               0 1 0 0 0 0
## PUDUCHERRY                            0 1 0 0 0 0
## PUNJAB                                    0 0 0 0 1 0
## RAJASTHAN                                0 0 1 0 0 0
## TAMIL NADU                               0 0 1 0 0 0
## TRIPURA                                  0 0 1 0 0 0

```

waterQuality_2017_Avg

```

##                                     State TemperatureAvg DissolvedO2Avg
## 1                               ANDHRA PRADESH      26.00000   7.100000
## 2 DAMAN AND DIU, DADRA\r\nAND NAGAR HAVELI 26.11111   6.816667
## 3                               GOA            26.75217   6.458696
## 4                               GUJARAT        24.58824   5.114706
## 5                               HARYANA        22.98681   4.440000
## 6                               HIMACHAL PRADESH 14.00000   6.441667
## 7                               KARNATAKA     27.00000   6.808333
## 8                               KERALA         25.71690   5.888732
## 9                               MAHARASHTRA    23.50000   6.504492
## 10                             MANIPUR       21.25758   6.572727
## 11                             MEGHALAYA    14.51852   6.214815
## 12                             PUDUCHERRY  29.50000   7.900000
## 13                             PUNJAB        14.92308   3.208143
## 14                             RAJASTHAN    13.00000   5.216667
## 15                             TAMIL NADU   24.41667   6.862500
## 16                             TRIPURA       24.78846   6.461538
##          pHAvg ConductivityAvg BODAvg NitrateAvg FecalColiformAvg
## 1 7.200000        462.5000 1.775000 1.7725000   18.500000
## 2 8.005556        13533.6667 6.144444 1.4544444   480.888889
## 3 7.071739        15039.6739 2.486957 1.6443968 10291.000000
## 4 7.644118        9336.4118 11.038235 0.4597059   149.794118
## 5 7.320000        1469.9000 49.920000 1.3161265   7606.787747
## 6 7.350000        843.8333 14.633333 0.7758333   42.250000
## 7 7.283333        137.5000 2.758333 0.8466667   269.833333
## 8 6.950000        4915.1761 2.775352 1.6994366 15450.626761
## 9 7.416667        4500.4583 8.186111 1.2793368   69.041667
## 10 7.050000       328.5303 7.139394 1.3161265   6909.646245
## 11 6.907407       198.8148 9.918519 1.9396296   5736.814815
## 12 8.075000       5908.5000 9.338235 0.1925000   6909.646245
## 13 7.688462       887.0769 75.826923 1.6226923 17319.230769
## 14 8.033333       566.6667 1.400000 1.6866667   8.666667
## 15 7.629167       2651.2500 2.262500 0.1800000   66.958333

```

```

## 16 7.369231      135.3462  1.109615  0.5936538      192.500000
##   TotalColiformAvg      NO2      PM10      SO2      PM2.5 cluster
## 1    342.50000 17.562500 28.25000  7.125000 17.562500      3
## 2    785.33333 18.269553 51.77669  9.068159 26.843819      6
## 3   14351.82609 13.411765 59.06424  7.021524 21.260108      1
## 4    812.94118 24.562963 67.99259 16.120823 29.083951      6
## 5   14633.61503 24.576923 78.03846 12.346154 26.329325      2
## 6    108.58333  8.454545 28.45455  2.272727  8.454545      3
## 7    717.41667 17.102624 47.58049  7.824026 24.676111      3
## 8   17967.12676 12.891386 43.55600  4.900582 24.997756      1
## 9    165.95833 34.694938 67.94619 11.944126 35.157701      3
## 10   410.19697 24.231229 38.50000  7.254072 32.640613      4
## 11   12302.44444  9.928571 39.37500  6.562912 23.854899      2
## 12   8974.35839 11.500000 41.83333  6.500000 39.281226      2
## 13   40500.00000 17.832208 67.48132  7.546342 23.521703      5
## 14   25.33333 28.032680 99.91756  9.937908 32.468740      3
## 15   124.29167 18.269553 51.77669  9.068159 26.843819      3
## 16   322.69231 13.000000 55.75000 21.000000 24.250000      3

```

```
Result_2017= mean_by_group[order(mean_by_group$TotalColiformAvg, mean_by_group$BODAvg, mean_by_group$Di
```

```
Result_2017$Rank=c("A", "B", "C", "D", "E","F")
```

```
class(Result_2017$cluster)
```

```
## [1] "integer"
```

```
class(waterQuality_2017_Avg$cluster)
```

```
## [1] "integer"
```

```
WaterAir2017 = merge(waterQuality_2017_Avg, Result_2017[,c("cluster","Rank")], by = "cluster", all.x = T
```

```
waterQuality_2018=finalDataset[finalDataset$year == 2018,]
waterQuality_2018$TemperatureMin      =as.numeric(waterQuality_2018$TemperatureMin)
waterQuality_2018$TemperatureMax     =as.numeric(waterQuality_2018$TemperatureMax)
waterQuality_2018$DissolvedO2Min     =as.numeric(waterQuality_2018$DissolvedO2Min)
```

```
## Warning: NAs introduced by coercion
```

```
waterQuality_2018$DissolvedO2Max     =as.numeric(waterQuality_2018$DissolvedO2Max)
```

```
## Warning: NAs introduced by coercion
```

```
waterQuality_2018$pHMin              =as.numeric(waterQuality_2018$pHMin)
waterQuality_2018$pHMax              =as.numeric(waterQuality_2018$pHMax)
waterQuality_2018$ConductivityMin   =as.numeric(waterQuality_2018$ConductivityMin)
waterQuality_2018$ConductivityMax   =as.numeric(waterQuality_2018$ConductivityMax)
waterQuality_2018$BODMin             =as.numeric(waterQuality_2018$BODMin)
```

```

## Warning: NAs introduced by coercion

waterQuality_2018$BODMax      =as.numeric(waterQuality_2018$BODMax)
waterQuality_2018$NitrateMin   =as.numeric(waterQuality_2018$NitrateMin)

## Warning: NAs introduced by coercion

waterQuality_2018$NitrateMax    =as.numeric(waterQuality_2018$NitrateMax)
waterQuality_2018$FecalColiformMin =as.numeric(waterQuality_2018$FecalColiformMin)
waterQuality_2018$FecalColiformMax =as.numeric(waterQuality_2018$FecalColiformMax)
waterQuality_2018$TotalColiformMin =as.numeric(waterQuality_2018$TotalColiformMin)
waterQuality_2018$TotalColiformMax =as.numeric(waterQuality_2018$TotalColiformMax)
waterQuality_2018$NO2           =as.numeric(waterQuality_2018$NO2)
waterQuality_2018$PM10          =as.numeric(waterQuality_2018$PM10)

cols=c("TemperatureMin"
,"TemperatureMax"
,"DissolvedO2Min"
,"DissolvedO2Max"
,"pHMin"
,"pHMax"
,"ConductivityMin"
,"ConductivityMax"
,"BODMin"
,"BODMax"
,"NitrateMin"
,"NitrateMax"
,"FecalColiformMin"
,"FecalColiformMax"
,"TotalColiformMin"
,"TotalColiformMax"
,"NO2"
,"PM10"
,"SO2"
,"PM2.5")

waterQuality_2018_Avg=waterQuality_2018

# Calculate the average values for each variable
waterQuality_2018_Avg$TemperatureAvg = (waterQuality_2018_Avg$TemperatureMin + waterQuality_2018_Avg$TemperatureMax) / 2
waterQuality_2018_Avg$DissolvedO2Avg = (waterQuality_2018_Avg$DissolvedO2Min + waterQuality_2018_Avg$DissolvedO2Max) / 2
waterQuality_2018_Avg$pHAvg = (waterQuality_2018_Avg$pHMin + waterQuality_2018_Avg$pHMax) / 2
waterQuality_2018_Avg$ConductivityAvg = (waterQuality_2018_Avg$ConductivityMin + waterQuality_2018_Avg$ConductivityMax) / 2
waterQuality_2018_Avg$FecalColiformAvg = (waterQuality_2018_Avg$FecalColiformMin + waterQuality_2018_Avg$FecalColiformMax) / 2
waterQuality_2018_Avg$BODAvg = (waterQuality_2018_Avg$BODMin + waterQuality_2018_Avg$BODMax) / 2
waterQuality_2018_Avg$NitrateAvg = (waterQuality_2018_Avg$NitrateMin + waterQuality_2018_Avg$NitrateMax) / 2

waterQuality_2018_Avg$TotalColiformAvg = (waterQuality_2018_Avg$TotalColiformMin + waterQuality_2018_Avg$TotalColiformMax) / 2

waterQuality_2018_Avg = subset(waterQuality_2018_Avg,select=-c(TemperatureMin
, TemperatureMax
,DissolvedO2Min
))

```

```

,DissolvedO2Max
,pHMin
,pHMax
,ConductivityMin
,ConductivityMax
,BODMin
,BODMax
,NitrateMin
,NitrateMax
,FecalColiformMin
,FecalColiformMax
,TotalColiformMin
,TotalColiformMax))

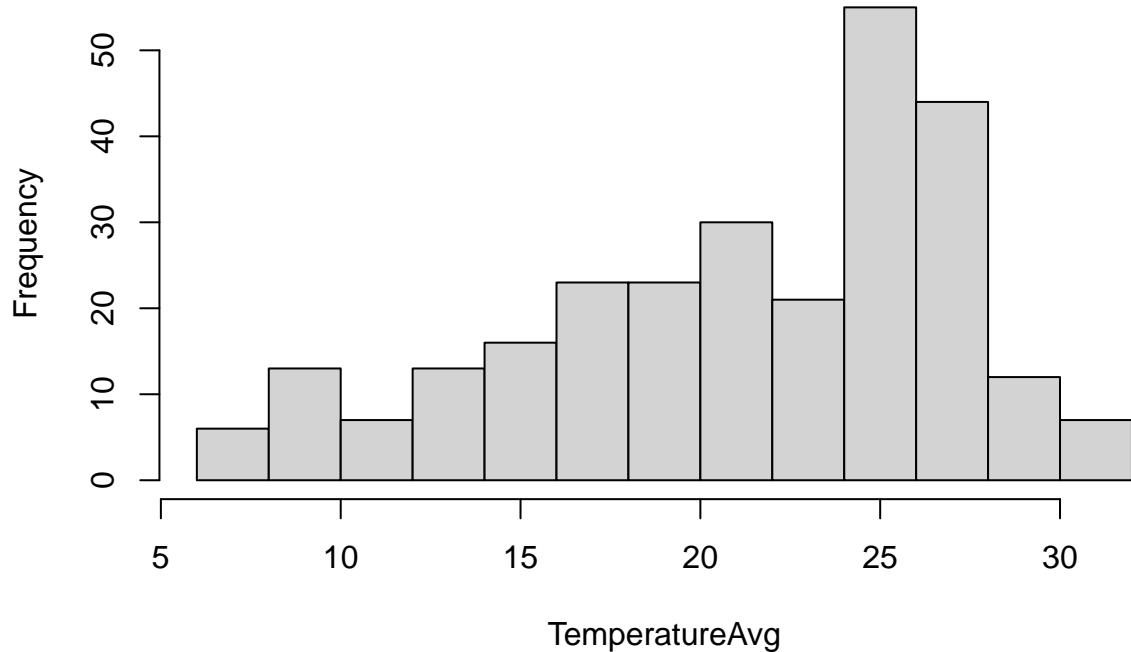
numCols=c("TemperatureAvg"
,"DissolvedO2Avg"
,"pHAvg"
,"ConductivityAvg"
,"BODAvg"
,"NitrateAvg"
,"TotalColiformAvg"
,"NO2"
,"PM10"
,"S02"
,"PM2.5")

for (col in colnames(waterQuality_2018_Avg)) {
  # Check if column has any NAs
  if (sum(is.na(waterQuality_2018_Avg[, col])) > 0) {
    # Replace NAs with mean value of the column
    waterQuality_2018_Avg[is.na(waterQuality_2018_Avg[, col]), col] = mean(waterQuality_2018_Avg[, col])
  }
}

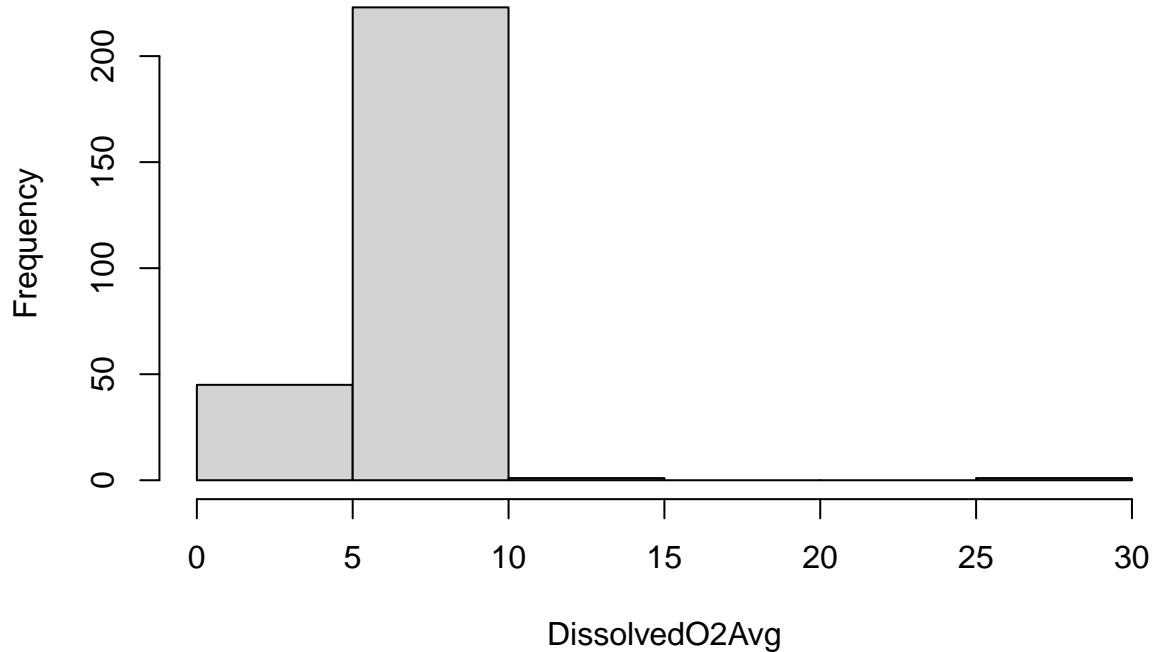
for (col in numCols) {
  # Create the histogram plot
  hist(waterQuality_2018_Avg[[col]], main = col, xlab = col)
}

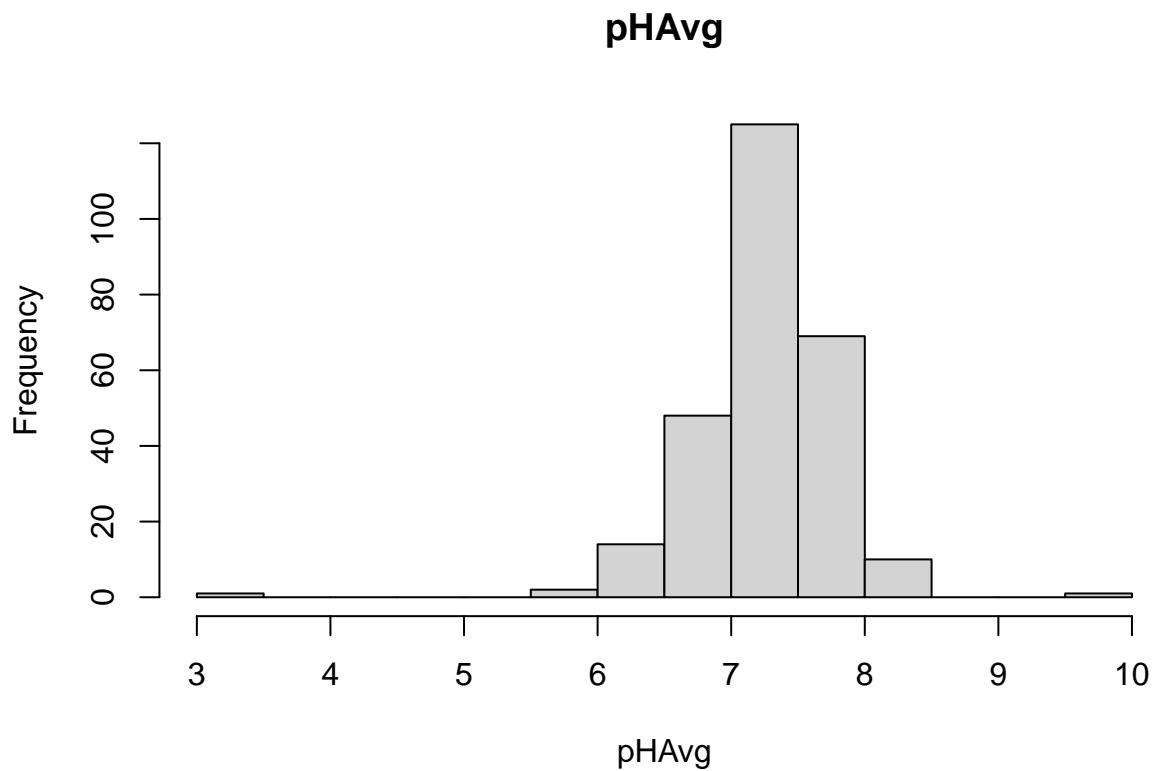
```

TemperatureAvg

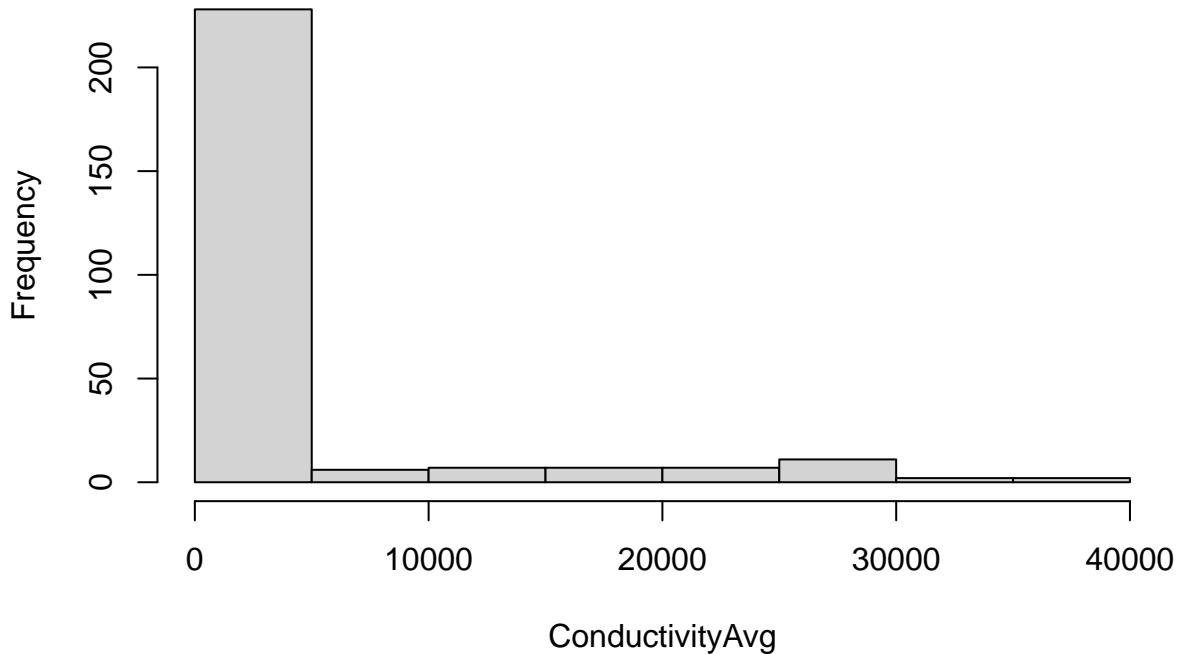


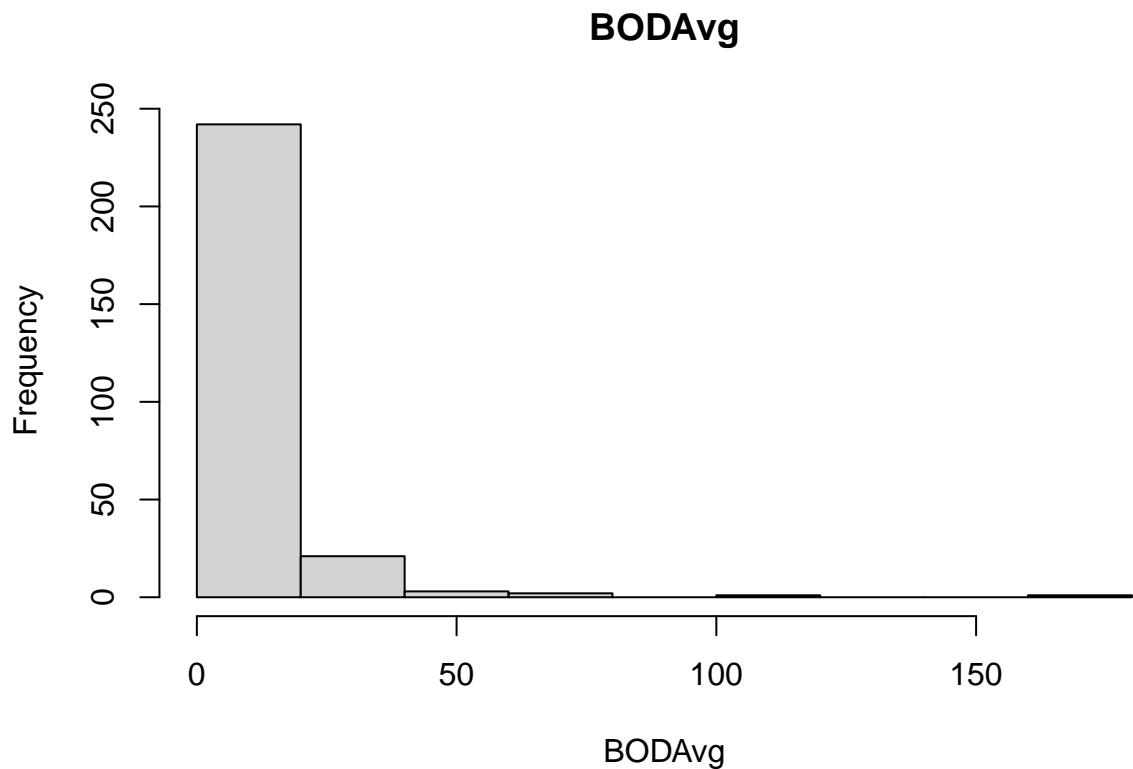
DissolvedO2Avg



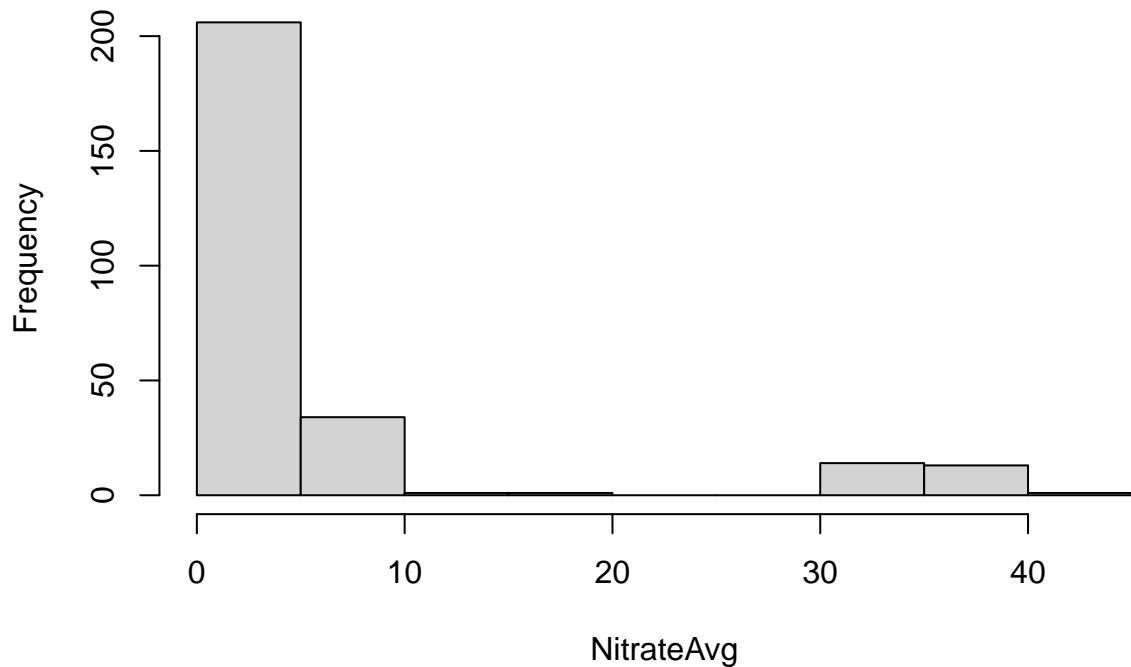


ConductivityAvg

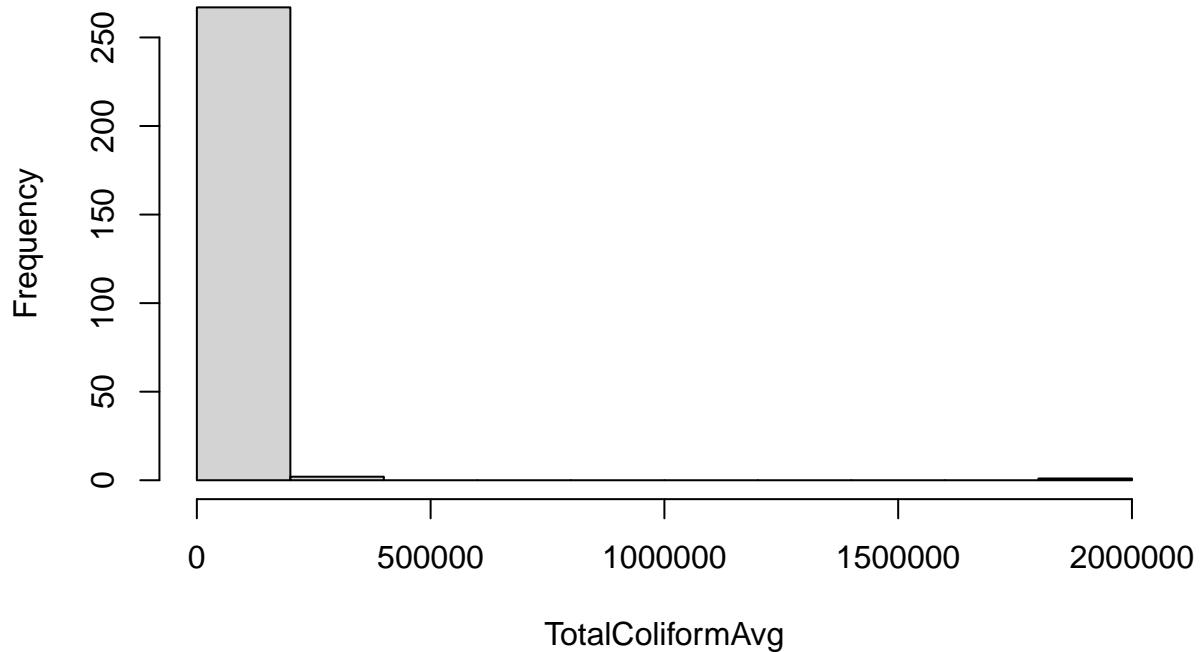




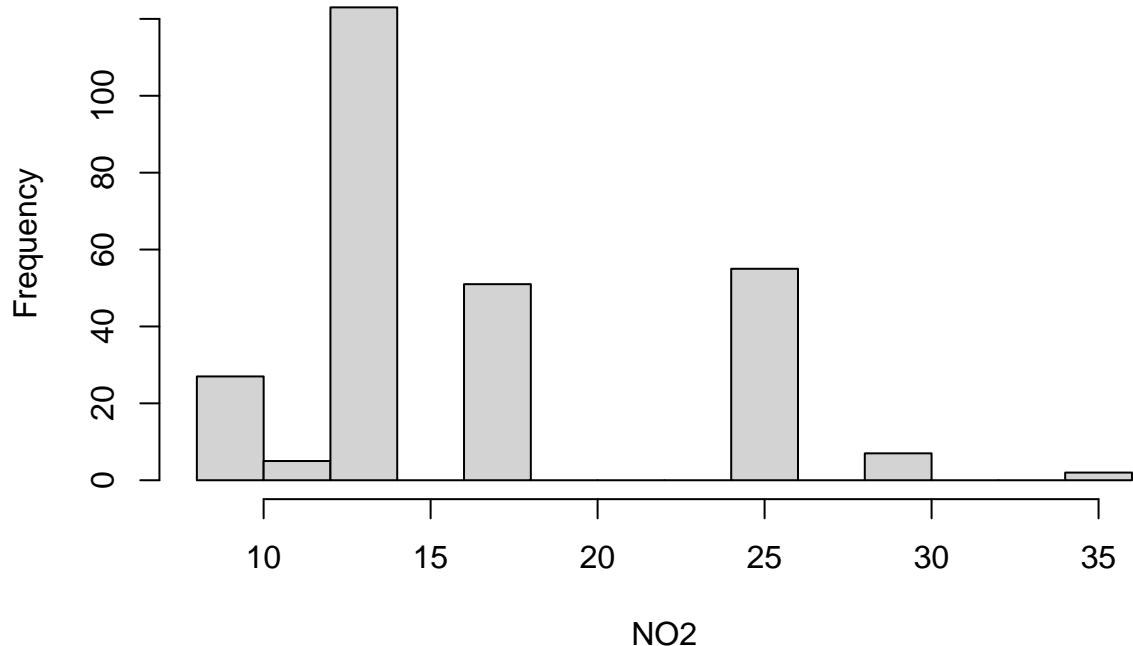
NitrateAvg

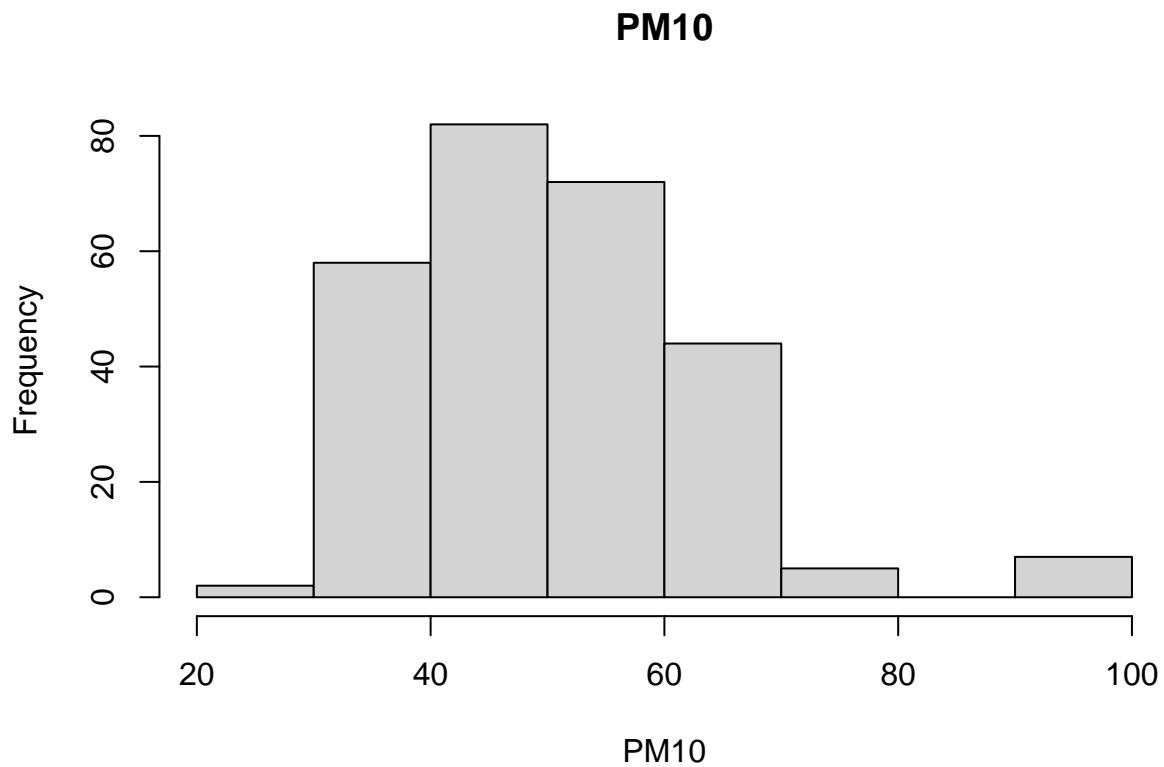


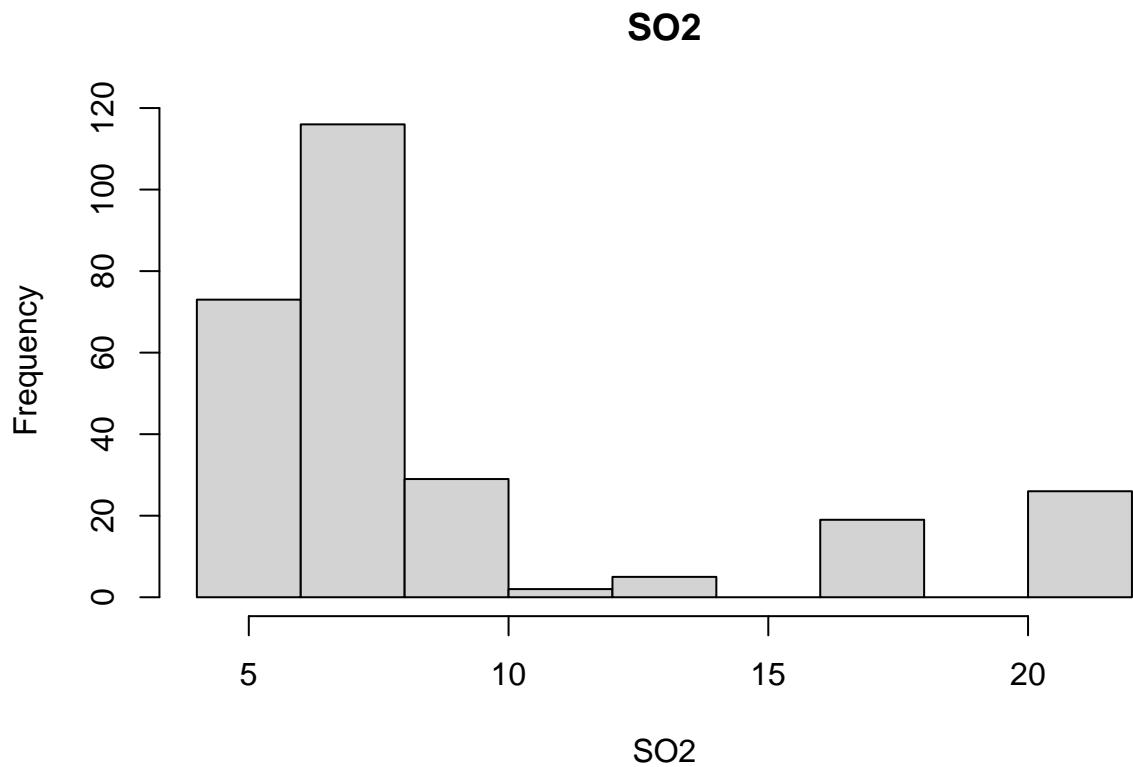
TotalColiformAvg

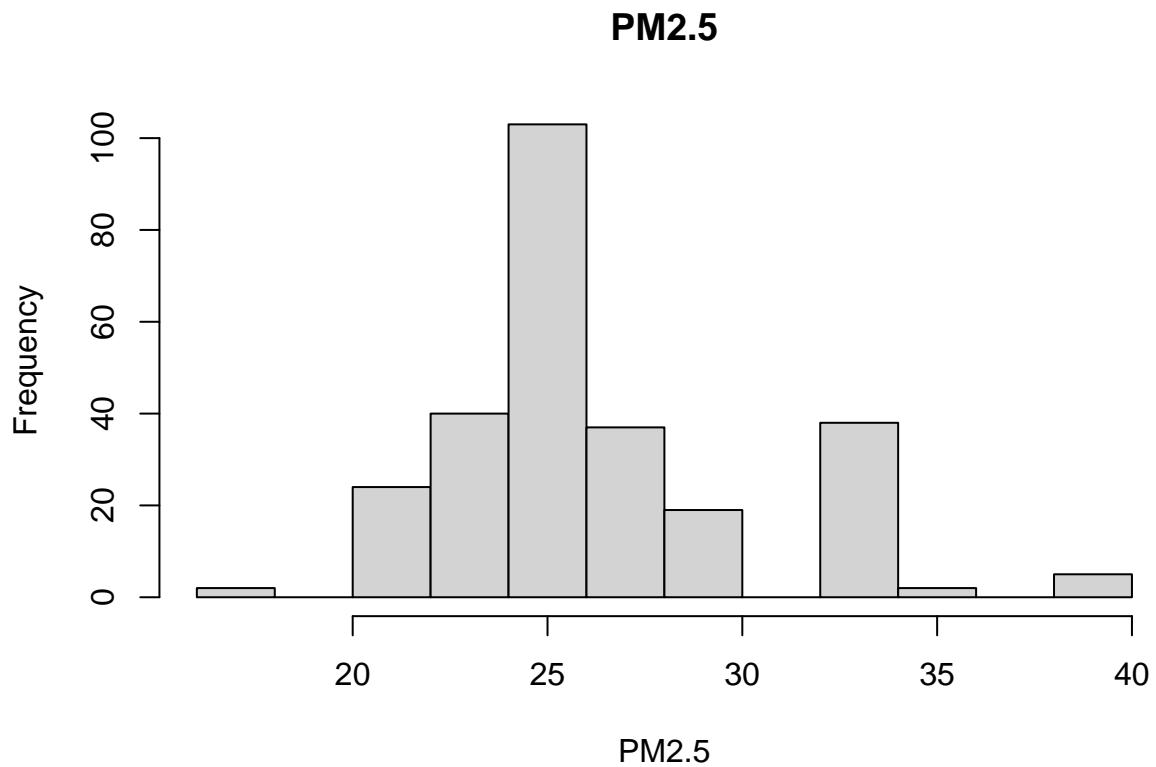


NO2

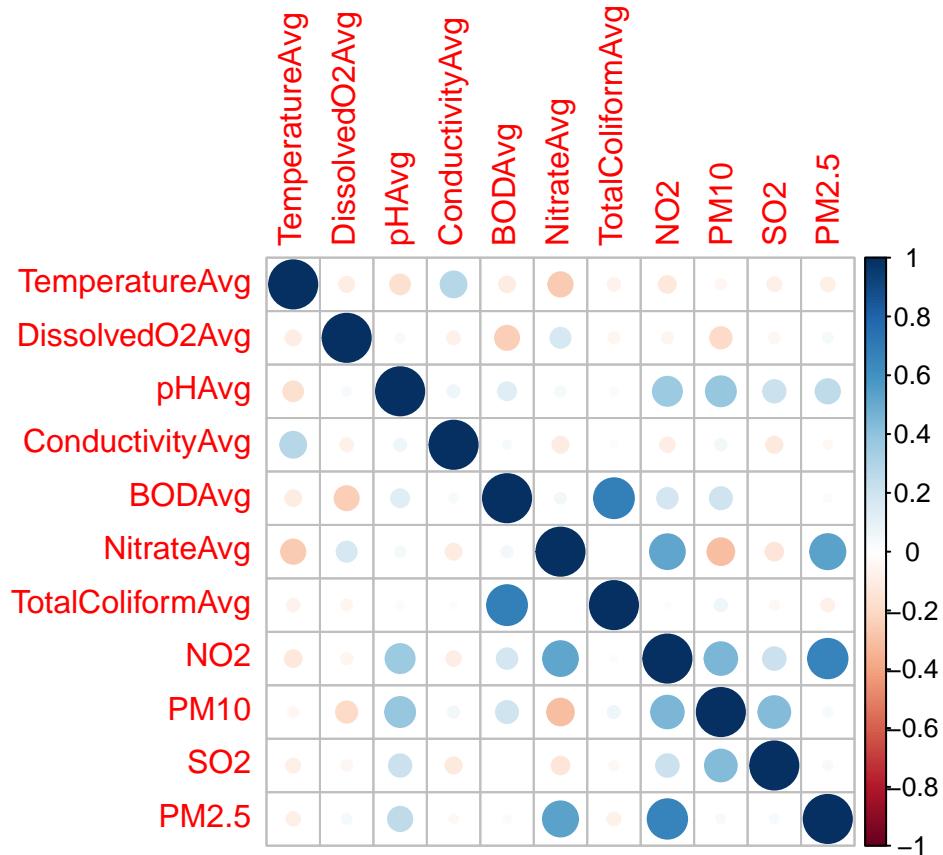








```
cov_mat = cov(waterQuality_2018_Avg[, numCols])  
  
corr_matrix = cor(waterQuality_2018_Avg[, numCols])  
  
library(corrplot)  
  
corrplot(corr_matrix)
```



```
waterQuality_2018_Avg = aggregate(cbind(TemperatureAvg,
                                         DissolvedO2Avg,
                                         pHAvg,
                                         ConductivityAvg,
                                         BODAvg,
                                         NitrateAvg,
                                         FecalColiformAvg,
                                         TotalColiformAvg,
                                         NO2,
                                         PM10,
                                         SO2,
                                         PM2.5) ~ State,
                                     data = waterQuality_2018_Avg,
                                     mean)
```

```
pca_data = waterQuality_2018_Avg[, c("TemperatureAvg"
                                         , "DissolvedO2Avg"
                                         , "pHAvg"
                                         , "ConductivityAvg"
                                         , "BODAvg"
                                         , "NitrateAvg"
                                         , "FecalColiformAvg"
                                         , "TotalColiformAvg"
                                         , "NO2"
                                         , "PM10"
                                         , "SO2")]
```

```

,"PM2.5")]

# Center and scale the data
pca_data_scaled = scale(pca_data)

# Perform PCA
pca_result = princomp(pca_data_scaled, center = TRUE, scale. = TRUE)

## Warning: In princomp.default(pca_data_scaled, center = TRUE, scale. = TRUE) :
## extra arguments 'center', 'scale.' will be disregarded

# Print the summary of PCA results
summary(pca_result)

## Importance of components:
##                               Comp.1    Comp.2    Comp.3    Comp.4    Comp.5
## Standard deviation      1.795372 1.6209385 1.2898093 1.1508861 0.84983365
## Proportion of Variance 0.286521 0.2335504 0.1478763 0.1177368 0.06419709
## Cumulative Proportion  0.286521 0.5200714 0.6679477 0.7856844 0.84988152
##                               Comp.6    Comp.7    Comp.8    Comp.9    Comp.10
## Standard deviation      0.78797250 0.70555331 0.5053609 0.45984654 0.281318668
## Proportion of Variance 0.05519117 0.04424937 0.0227013 0.01879634 0.007034684
## Cumulative Proportion  0.90507269 0.94932207 0.9720234 0.99081971 0.997854392
##                               Comp.11   Comp.12
## Standard deviation      0.154473980 1.660946e-02
## Proportion of Variance 0.002121085 2.452215e-05
## Cumulative Proportion  0.999975478 1.000000e+00

#reverse the signs
pca_result$rotation = -1*pca_result$rotation

#display principal components
pca_result$rotation

## numeric(0)

var_explained = pca_result$sdev^2 / sum(pca_result$sdev^2)

print("Before")

## [1] "Before"

print(var_explained)

##                               Comp.1    Comp.2    Comp.3    Comp.4    Comp.5    Comp.6
## 2.865210e-01 2.335504e-01 1.478763e-01 1.177368e-01 6.419709e-02 5.519117e-02
##                               Comp.7    Comp.8    Comp.9    Comp.10   Comp.11   Comp.12
## 4.424937e-02 2.270130e-02 1.879634e-02 7.034684e-03 2.121085e-03 2.452215e-05

```

```

pc.comp = pca_result$scores

#86%
pc.comp1 = -1*pc.comp[,1]
pc.comp2 = -1*pc.comp[,2]
pc.comp3 = -1*pc.comp[,3]
pc.comp4 = -1*pc.comp[,4]

library(factoextra)
library(tidyr)
library(ggplot2)

X = cbind(pc.comp1, pc.comp2, pc.comp3, pc.comp4)

X

##          pc.comp1   pc.comp2   pc.comp3   pc.comp4
## [1,] -1.8979617  0.9787115  0.3890051  0.6192182
## [2,] -0.1017835 -0.7234323 -1.4084539  0.1405186
## [3,] -1.2430553  0.5481157 -1.4245127  0.3465032
## [4,]  1.2785995 -2.0569369 -0.2674963  0.8396035
## [5,]  2.7997885 -1.4014173  0.0513338  0.5135111
## [6,] -1.6751929  0.5227284  0.5111217  0.3094810
## [7,] -1.9359276  0.8591319 -0.6030710  0.5237516
## [8,]  0.1228824 -2.0967510  0.9813471 -0.8006504
## [9,] -1.0335265  0.5769160  2.4711766 -3.1266757
## [10,] -0.5638875  2.1482111  0.5660710  0.7952671
## [11,] -0.2201336  0.2388891  0.6956469 -0.2286696
## [12,] -1.0843620 -0.8343389 -3.4160246 -1.8574002
## [13,]  4.5184202  4.1093009 -0.5978607 -0.4048123
## [14,]  2.5024650 -2.4437442  0.6746147 -0.1393086
## [15,] -0.6653201 -0.1618106  0.2140996  0.3556425
## [16,] -0.8010048 -0.2635735  1.1630027  2.1140198

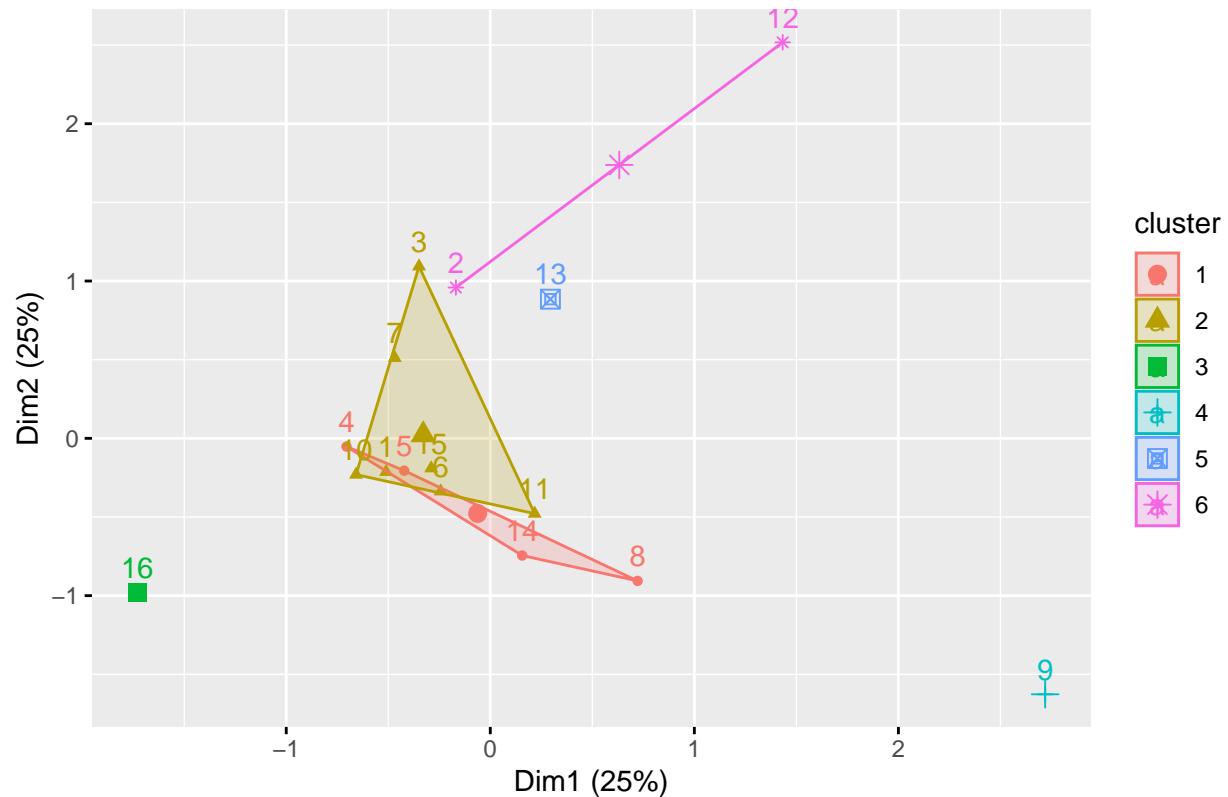
k = 6
kmeans_model = kmeans(X, centers = k, nstart=1)

waterQuality_2018_Avg$cluster=kmeans_model$cluster

fviz_cluster(kmeans_model, data = X)

```

Cluster plot



```
mean_by_group = aggregate(cbind(TemperatureAvg, DissolvedO2Avg, pHAvg, ConductivityAvg, BODAvg, NitrateAvg),
                          by = list(waterQuality_2018_Avg$State), FUN = mean)
table(waterQuality_2018_Avg$State, waterQuality_2018_Avg$cluster)
```

```
##                                     1 2 3 4 5 6
## ANDHRA PRADESH                 0 1 0 0 0 0
## DAMAN, DIU, DADRA\r\nNAGAR HAVELI 0 0 0 0 0 1
## GOA                            0 1 0 0 0 0
## GUJARAT                         1 0 0 0 0 0
## HARYANA                          1 0 0 0 0 0
## KARNATAKA                        0 1 0 0 0 0
## KERALA                           0 1 0 0 0 0
## MAHARASHTRA                      1 0 0 0 0 0
## MANIPUR                           0 0 0 1 0 0
## MEGHALAYA                         0 1 0 0 0 0
## ODISHA                            0 1 0 0 0 0
## PUDUCHERRY                        0 0 0 0 0 1
## PUNJAB                            0 0 0 0 1 0
## RAJASTHAN                          1 0 0 0 0 0
## TAMIL NADU                         0 1 0 0 0 0
## TRIPURA                           0 0 1 0 0 0
```

```
waterQuality_2018_Avg
```

```

##                                     State TemperatureAvg DissolvedO2Avg      pHAvg
## 1          ANDHRA PRADESH           25.00000  6.175000 7.225000
## 2  DAMAN, DIU, DADRA\r\nNAGAR HAVELI  28.06667  5.738889 7.755556
## 3                      GOA           26.13333  6.264583 7.204167
## 4                      GUJARAT        23.87576  4.634211 7.671053
## 5                      HARYANA        21.60656  4.200000 7.880000
## 6                      KARNATAKA       25.75000  7.125000 7.150000
## 7                      KERALA           26.00000  6.039726 6.928082
## 8                      MAHARASHTRA      27.50000  6.750000 7.450000
## 9                      MANIPUR          17.17419  7.280645 7.322581
## 10                     MEGHALAYA        12.07778  6.175926 7.135185
## 11                     ODISHA            14.80000  7.195000 7.620000
## 12                     PUDUCHERRY       30.40000  6.480000 7.690000
## 13                     PUNJAB            14.15385  6.073771 7.561538
## 14                     RAJASTHAN          15.91429  4.728571 7.950000
## 15                     TAMIL NADU         23.45385  6.580769 7.588462
## 16                     TRIPURA           20.28846  6.688462 7.226923
##   ConductivityAvg      BODAvg NitrateAvg FecalColiformAvg TotalColiformAvg
## 1     2348.2500  1.5250000  8.5425000    34.00000  957.5000
## 2     7339.8333 13.0388889  0.9705556   321.50000  713.5000
## 3    12217.9792  4.1847707  4.5083412   6313.35417 13139.5417
## 4     5579.0263 11.8815789  0.7110526   287.42105  647.7632
## 5    1580.0000 28.1200000  5.1475105   5395.43707 12308.3760
## 6     108.2500  2.3625000  2.9862552   240.12500  863.7500
## 7    4282.7260  2.1691781  1.0863704   1509.34932 2773.3630
## 8     242.5000  3.2500000  0.9000000   23.50000  70.5000
## 9     628.5000  7.7290323  33.2595168   6992.70495 207.9677
## 10    231.2037  7.1685185  0.7035185  12092.51852 31783.9074
## 11    2132.1500  1.3400000  1.5660000   4463.55000  7547.0000
## 12   17812.4000  7.4693798  0.2540000   7225.72845 14463.9599
## 13   1073.6154 38.6923077  1.4123077  78800.00000 175400.0000
## 14   2189.2857  7.9285714  2.3735714   19.07143  67.0000
## 15   253.2308  3.8115385  0.8826923   32.30769  108.5769
## 16   122.4423  0.6115385  0.3880769   147.96154  276.9231
##   NO2      PM10      S02      PM2.5 cluster
## 1  17.562500 28.25000  7.125000 17.56250  2
## 2  16.294960 51.56409  8.801921 26.26899  6
## 3  13.411765 59.06424  7.021524 21.26011  2
## 4  24.562963 67.99259 16.120823 29.08395  1
## 5  24.576923 78.03846 12.346154 26.32933  1
## 6  17.102624 47.58049  7.824026 24.67611  2
## 7  12.891386 43.55600  4.900582 24.99776  2
## 8  34.694938 67.94619 11.944126 35.15770  1
## 9  24.231229 38.50000  7.254072 32.64061  4
## 10  9.928571 39.37500  6.562912 23.85490  2
## 11 17.290323 63.40261  7.091398 27.53333  2
## 12 11.500000 41.83333  6.500000 39.28123  6
## 13 17.832208 67.48132  7.546342 23.52170  5
## 14 28.032680 99.91756  9.937908 32.46874  1
## 15 16.294960 51.56409  8.801921 26.26899  2
## 16 13.000000 55.75000 21.000000 24.25000  3

```

```
Result_2018= mean_by_group[order(mean_by_group$TotalColiformAvg, mean_by_group$BODAvg, mean_by_group$Di
```

```

Result_2018$Rank=c("A", "B", "C", "D", "E","F")

class(Result_2018$cluster)

## [1] "integer"

class(waterQuality_2018_Avg$cluster)

## [1] "integer"

WaterAir2018 = merge(waterQuality_2018_Avg, Result_2018[,c("cluster","Rank")], by = "cluster", all.x = T)

WaterQualityAlone = waterQuality_2018_Avg[, c("TemperatureAvg"
,"DissolvedO2Avg"
,"pHAvg"
,"ConductivityAvg"
,"BODAvg"
,"FecalColiformAvg"
,"NitrateAvg"
,"TotalColiformAvg"
)] 

# Center and scale the data
WaterQualityAlone_scaled = scale(WaterQualityAlone)

# Perform PCA
pca_result = princomp(WaterQualityAlone_scaled, center = TRUE, scale. = TRUE)

## Warning: In princomp.default(WaterQualityAlone_scaled, center = TRUE, scale. = TRUE) :
##   extra arguments 'center', 'scale.' will be disregarded

# Print the summary of PCA results
summary(pca_result)

## Importance of components:
##                               Comp.1    Comp.2    Comp.3    Comp.4    Comp.5
## Standard deviation     1.7022634 1.3441246 1.0773239 0.9073936 0.66410055
## Proportion of Variance 0.3863601 0.2408895 0.1547502 0.1097817 0.05880394
## Cumulative Proportion  0.3863601 0.6272495 0.7819998 0.8917815 0.95058546
##                               Comp.6    Comp.7    Comp.8
## Standard deviation     0.57368627 0.202426610 2.272925e-02
## Proportion of Variance 0.04388212 0.005463538 6.888253e-05
## Cumulative Proportion  0.99446758 0.999931117 1.000000e+00

#reverse the signs
pca_result$rotation = -1*pca_result$rotation

#display principal components
pca_result$rotation

```

```

## numeric(0)

var_explained = pca_result$sdev^2 / sum(pca_result$sdev^2)
print("After")

## [1] "After"

print(var_explained)

##          Comp.1      Comp.2      Comp.3      Comp.4      Comp.5      Comp.6
## 3.863601e-01 2.408895e-01 1.547502e-01 1.097817e-01 5.880394e-02 4.388212e-02
##          Comp.7      Comp.8
## 5.463538e-03 6.888253e-05

pc.comp = pca_result$scores

#86%
pc.comp1 = -1*pc.comp[,1]
pc.comp2 = -1*pc.comp[,2]
pc.comp3 = -1*pc.comp[,3]
pc.comp4 = -1*pc.comp[,4]
pc.comp5 = -1*pc.comp[,5]
pc.comp6 = -1*pc.comp[,6]

library(factoextra)
library(tidyr)
library(ggplot2)

X = cbind(pc.comp1, pc.comp2, pc.comp3, pc.comp4, pc.comp5, pc.comp6)

X

##          pc.comp1      pc.comp2      pc.comp3      pc.comp4      pc.comp5      pc.comp6
## [1,] -1.08440520  0.54357031 -0.02785999 -0.15234880 -0.65409629 -0.04198959
## [2,] -0.31493332 -1.70297492 -0.13844562 -0.36220550  0.11031889 -0.52426251
## [3,] -0.93635006 -0.43789029 -1.51468621 -0.57132265 -0.23341311  0.82893934
## [4,]  0.07102067 -1.76438229  0.70155024  0.08847368 -0.49331280  0.31181732
## [5,]  1.57784224 -1.85925455  1.72304365 -0.42166877 -0.85117490 -0.27330121
## [6,] -1.24019860  1.08060776 -0.52113444  0.54452799 -0.29846523 -0.69831086
## [7,] -1.22207214  0.38385853 -0.88646172  0.66044466 -1.19120550  0.42156637
## [8,] -1.02854966  0.15432140 -0.18182365  0.51982091  0.02593051 -1.09801080
## [9,] -0.40248614  2.70949254  1.05073326 -2.72557547 -0.04193558  0.01417763
## [10,]  0.65737064  1.41512599  0.35336142  1.14367368 -0.03028694  1.10174786
## [11,] -0.21339445  0.89349144  0.45962986  0.51887168  1.63241117  0.18235004
## [12,] -0.80040834 -2.08058276 -2.03981520 -1.08205093  0.91038568  0.24292782
## [13,]  5.90497553  0.83450638 -1.40882229  0.05358687 -0.03210360 -0.29653687
## [14,]  0.54866250 -1.22348226  1.99628722  0.26534724  0.59597581  0.55053427
## [15,] -0.64364806  0.08451558  0.35682748  0.53827538  0.47691957 -0.74906869
## [16,] -0.87342563  0.96907715  0.07761600  0.98215003  0.07405233  0.02741988

```

```

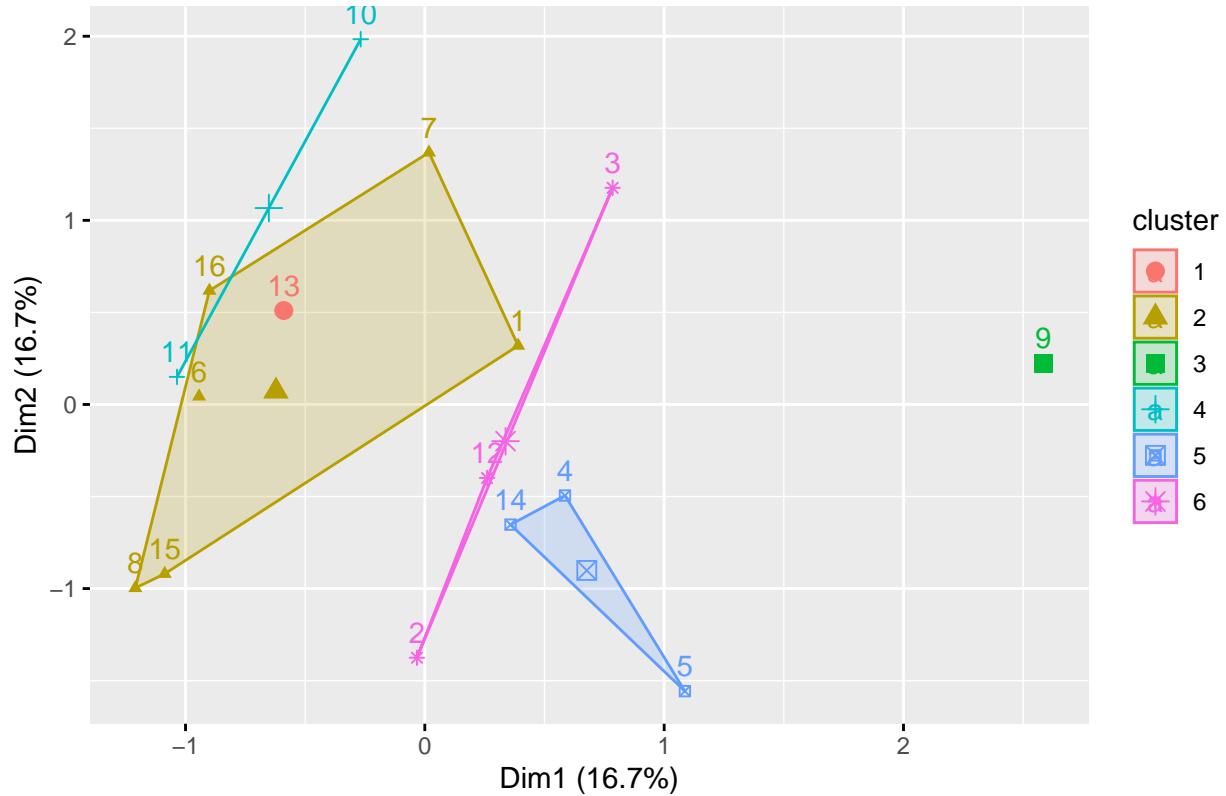
k = 6
kmeans_model = kmeans(X, centers = k, nstart=1)

waterQuality_2018_Avg$cluster=kmeans_model$cluster

fviz_cluster(kmeans_model, data = X)

```

Cluster plot



```
mean_by_group = aggregate(cbind(TemperatureAvg, DissolvedO2Avg, pHAvg, ConductivityAvg, BODAvg, NitrateAvg)
```

```
table(waterQuality_2018_Avg$State, waterQuality_2018_Avg$cluster)
```

```

##                                     1 2 3 4 5 6
## ANDHRA PRADESH                  0 1 0 0 0 0
## DAMAN , DIU, DADRA\r\nNAGAR HAVELI 0 0 0 0 0 1
## GOA                           0 0 0 0 0 1
## GUJARAT                      0 0 0 0 1 0
## HARYANA                       0 0 0 0 1 0
## KARNATAKA                     0 1 0 0 0 0
## KERALA                        0 1 0 0 0 0
## MAHARASHTRA                   0 1 0 0 0 0
## MANIPUR                       0 0 1 0 0 0
## MEGHALAYA                     0 0 0 1 0 0

```

```

##    ODISHA          0 0 0 1 0 0
##    PUDUCHERRY     0 0 0 0 0 1
##    PUNJAB          1 0 0 0 0 0
##    RAJASTHAN        0 0 0 0 1 0
##    TAMIL NADU      0 1 0 0 0 0
##    TRIPURA          0 1 0 0 0 0

```

WaterQualityAlone

```

##    TemperatureAvg DissolvedO2Avg pHAvg ConductivityAvg BODAvg
## 1      25.00000    6.175000 7.225000    2348.2500 1.5250000
## 2      28.06667    5.738889 7.755556    7339.8333 13.0388889
## 3      26.13333    6.264583 7.204167   12217.9792  4.1847707
## 4      23.87576    4.634211 7.671053    5579.0263 11.8815789
## 5      21.60656    4.200000 7.880000    1580.0000 28.1200000
## 6      25.75000    7.125000 7.150000     108.2500 2.3625000
## 7      26.00000    6.039726 6.928082   4282.7260 2.1691781
## 8      27.50000    6.750000 7.450000     242.5000 3.2500000
## 9      17.17419    7.280645 7.322581     628.5000 7.7290323
## 10     12.07778    6.175926 7.135185     231.2037 7.1685185
## 11     14.80000    7.195000 7.620000   2132.1500 1.3400000
## 12     30.40000    6.480000 7.690000   17812.4000 7.4693798
## 13     14.15385    6.073771 7.561538   1073.6154 38.6923077
## 14     15.91429    4.728571 7.950000   2189.2857 7.9285714
## 15     23.45385    6.580769 7.588462   253.2308 3.8115385
## 16     20.28846    6.688462 7.226923   122.4423 0.6115385
##    FecalColiformAvg NitrateAvg TotalColiformAvg
## 1      34.00000   8.5425000 957.5000
## 2      321.50000  0.9705556 713.5000
## 3      6313.35417 4.5083412 13139.5417
## 4      287.42105  0.7110526 647.7632
## 5      5395.43707 5.1475105 12308.3760
## 6      240.12500  2.9862552 863.7500
## 7      1509.34932 1.0863704 2773.3630
## 8      23.50000  0.9000000 70.5000
## 9      6992.70495 33.2595168 207.9677
## 10     12092.51852 0.7035185 31783.9074
## 11     4463.55000 1.5660000 7547.0000
## 12     7225.72845 0.2540000 14463.9599
## 13     78800.00000 1.4123077 175400.0000
## 14     19.07143   2.3735714 67.0000
## 15     32.30769   0.8826923 108.5769
## 16     147.96154  0.3880769 276.9231

```

```
Result_2018_water<- mean_by_group[order(mean_by_group$TotalColiformAvg, mean_by_group$BODAvg, mean_by_group$FecalColiformAvg),]
```

```
Result_2018_water$Rank=c("A", "B", "C", "D", "E", "F")
```

```
class(waterQuality_2018_Avg$cluster)
```

```
## [1] "integer"
```

```

class(waterQuality_2018_Avg$cluster)

## [1] "integer"

Water2018 = merge(waterQuality_2018_Avg, Result_2018_water[,c("cluster","Rank")], by = "cluster", all.x = TRUE)

waterQuality_2019=finalDataset[finalDataset$year == 2019,]
waterQuality_2019$TemperatureMin      =as.numeric(waterQuality_2019$TemperatureMin)
waterQuality_2019$TemperatureMax     =as.numeric(waterQuality_2019$TemperatureMax)
waterQuality_2019$DissolvedO2Min    =as.numeric(waterQuality_2019$DissolvedO2Min)

## Warning: NAs introduced by coercion

waterQuality_2019$DissolvedO2Max      =as.numeric(waterQuality_2019$DissolvedO2Max)

## Warning: NAs introduced by coercion

waterQuality_2019$pHMin              =as.numeric(waterQuality_2019$pHMin)
waterQuality_2019$pHMax              =as.numeric(waterQuality_2019$pHMax)
waterQuality_2019$ConductivityMin    =as.numeric(waterQuality_2019$ConductivityMin)
waterQuality_2019$ConductivityMax   =as.numeric(waterQuality_2019$ConductivityMax)
waterQuality_2019$BODMin             =as.numeric(waterQuality_2019$BODMin)

## Warning: NAs introduced by coercion

waterQuality_2019$BODMax             =as.numeric(waterQuality_2019$BODMax)

## Warning: NAs introduced by coercion

waterQuality_2019$NitrateMin         =as.numeric(waterQuality_2019$NitrateMin)

## Warning: NAs introduced by coercion

waterQuality_2019$NitrateMax         =as.numeric(waterQuality_2019$NitrateMax)
waterQuality_2019$FecalColiformMin   =as.numeric(waterQuality_2019$FecalColiformMin)

## Warning: NAs introduced by coercion

waterQuality_2019$FecalColiformMax   =as.numeric(waterQuality_2019$FecalColiformMax)
waterQuality_2019$TotalColiformMin   =as.numeric(waterQuality_2019$TotalColiformMin)
waterQuality_2019$TotalColiformMax   =as.numeric(waterQuality_2019$TotalColiformMax)
waterQuality_2019$N02                =as.numeric(waterQuality_2019$N02)
waterQuality_2019$PM10               =as.numeric(waterQuality_2019$PM10)

cols=c("TemperatureMin"
,"TemperatureMax"
,"DissolvedO2Min"
)

```

```

,"DissolvedO2Max"
,"pHMin"
,"pHMax"
,"ConductivityMin"
,"ConductivityMax"
,"BODMin"
,"BODMax"
,"NitrateMin"
,"NitrateMax"
,"FecalColiformMin"
,"FecalColiformMax"
,"TotalColiformMin"
,"TotalColiformMax"
,"NO2"
,"PM10"
,"SO2"
,"PM2.5")

waterQuality_2019_Avg=waterQuality_2019

# Calculate the average values for each variable
waterQuality_2019_Avg$TemperatureAvg = (waterQuality_2019_Avg$TemperatureMin + waterQuality_2019_Avg$TemperatureMax) / 2
waterQuality_2019_Avg$DissolvedO2Avg = (waterQuality_2019_Avg$DissolvedO2Min + waterQuality_2019_Avg$DissolvedO2Max) / 2
waterQuality_2019_Avg$pHAvg = (waterQuality_2019_Avg$pHMin + waterQuality_2019_Avg$pHMax) / 2
waterQuality_2019_Avg$ConductivityAvg = (waterQuality_2019_Avg$ConductivityMin + waterQuality_2019_Avg$ConductivityMax) / 2
waterQuality_2019_Avg$FecalColiformAvg = (waterQuality_2019_Avg$FecalColiformMin + waterQuality_2019_Avg$FecalColiformMax) / 2
waterQuality_2019_Avg$BODAvg = (waterQuality_2019_Avg$BODMin + waterQuality_2019_Avg$BODMax) / 2
waterQuality_2019_Avg$NitrateAvg = (waterQuality_2019_Avg$NitrateMin + waterQuality_2019_Avg$NitrateMax) / 2

waterQuality_2019_Avg$TotalColiformAvg = (waterQuality_2019_Avg$TotalColiformMin + waterQuality_2019_Avg$TotalColiformMax) / 2

waterQuality_2019_Avg = subset(waterQuality_2019_Avg, select=-c(TemperatureMin
, TemperatureMax
, DissolvedO2Min
, DissolvedO2Max
, pHMin
, pHMax
, ConductivityMin
, ConductivityMax
, BODMin
, BODMax
, NitrateMin
, NitrateMax
, FecalColiformMin
, FecalColiformMax
, TotalColiformMin
, TotalColiformMax))

numCols=c("TemperatureAvg"
,"DissolvedO2Avg"
,"pHAvg"
,"ConductivityAvg"
,
```

```

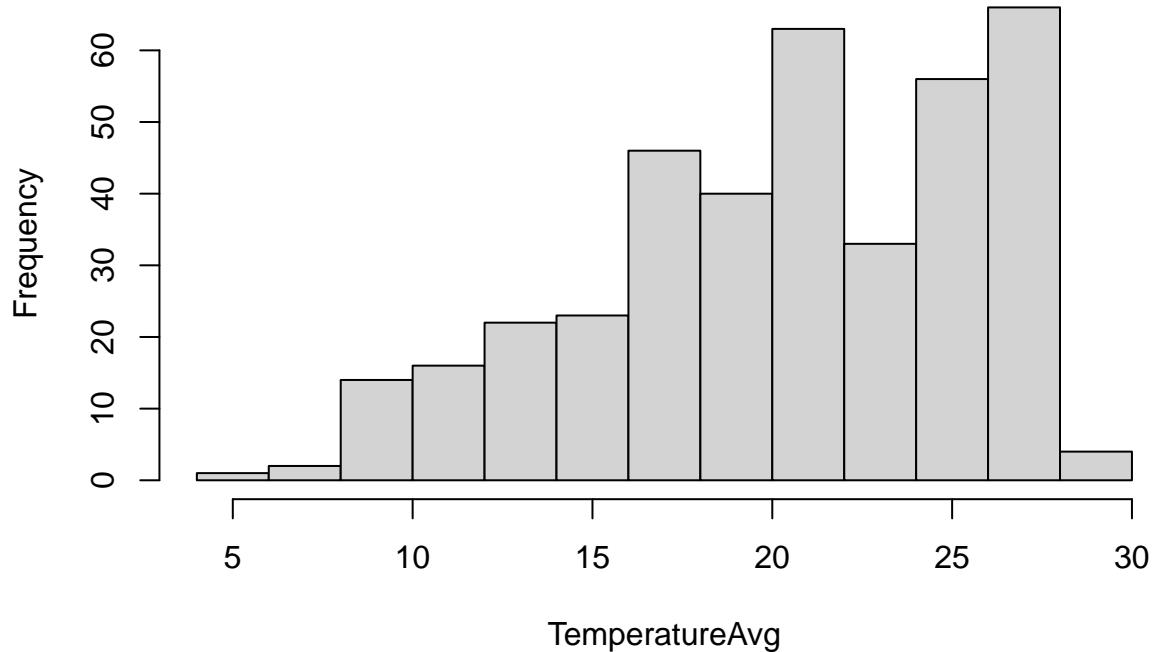
,"BODAvg"
,"FecalColiformAvg"
,"NitrateAvg"
,"TotalColiformAvg"
,"NO2"
,"PM10"
,"SO2"
,"PM2.5")

for (col in colnames(waterQuality_2019_Avg)) {
  # Check if column has any NAs
  if (sum(is.na(waterQuality_2019_Avg[, col])) > 0) {
    # Replace NAs with mean value of the column
    waterQuality_2019_Avg[is.na(waterQuality_2019_Avg[, col]), col] = mean(waterQuality_2019_Avg[, col])
  }
}

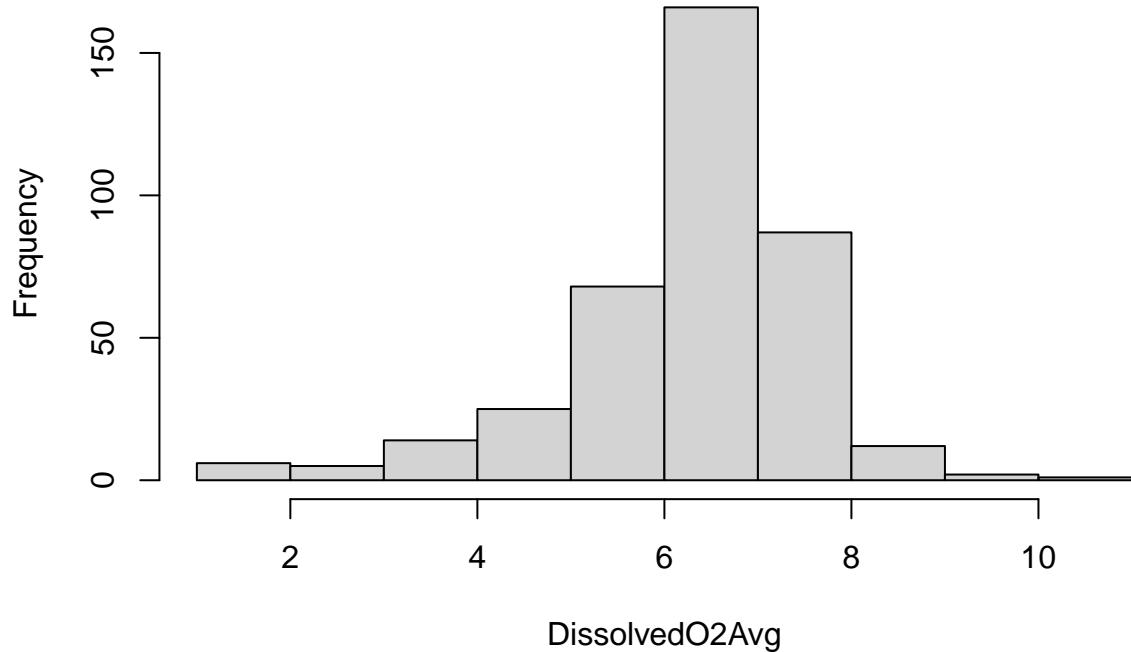
for (col in numCols) {
  # Create the histogram plot
  hist(waterQuality_2019_Avg[[col]], main = col, xlab = col)
}

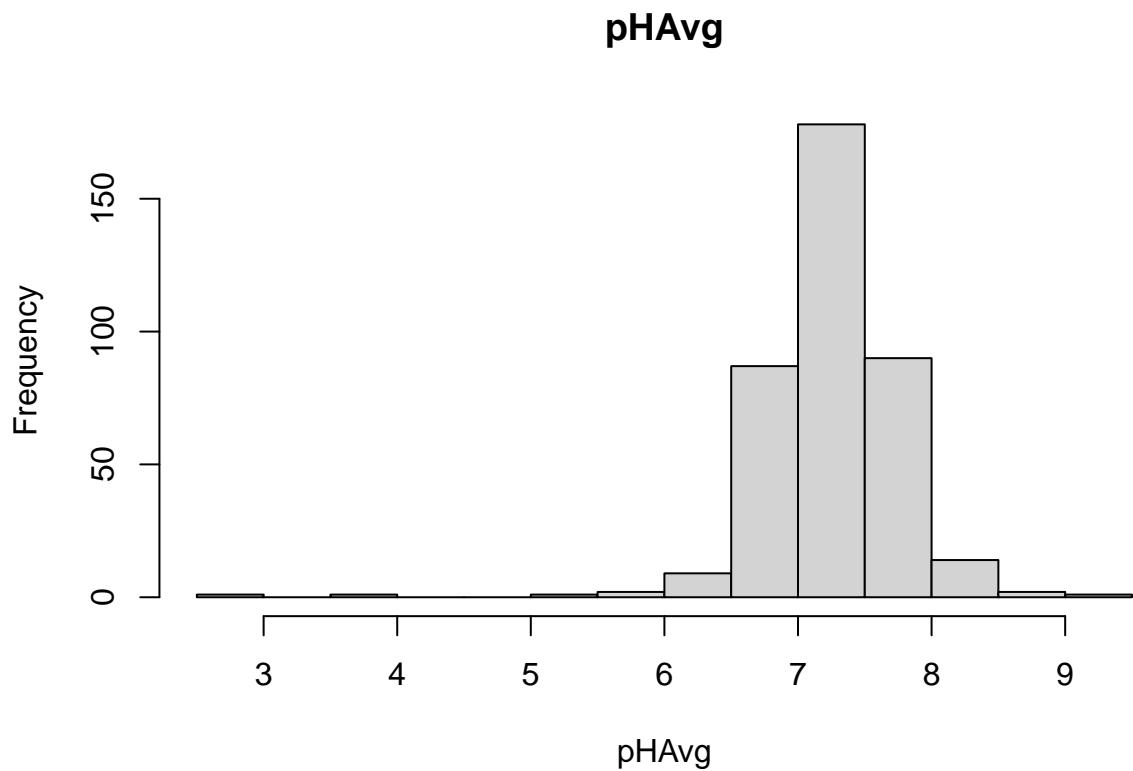
```

TemperatureAvg

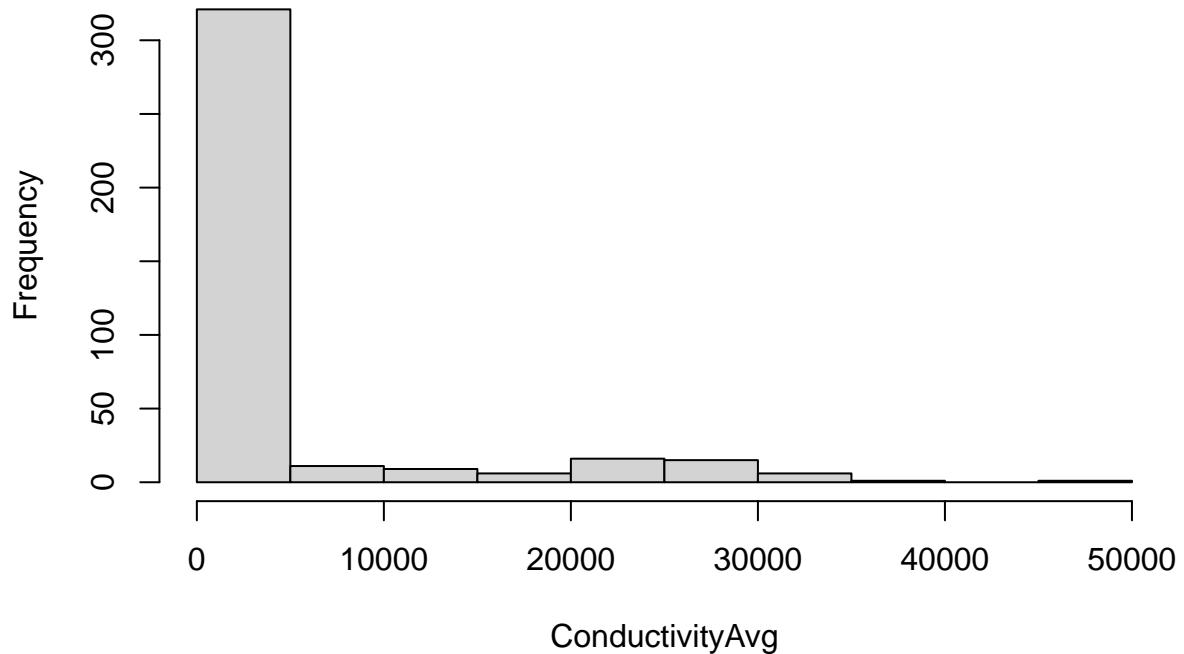


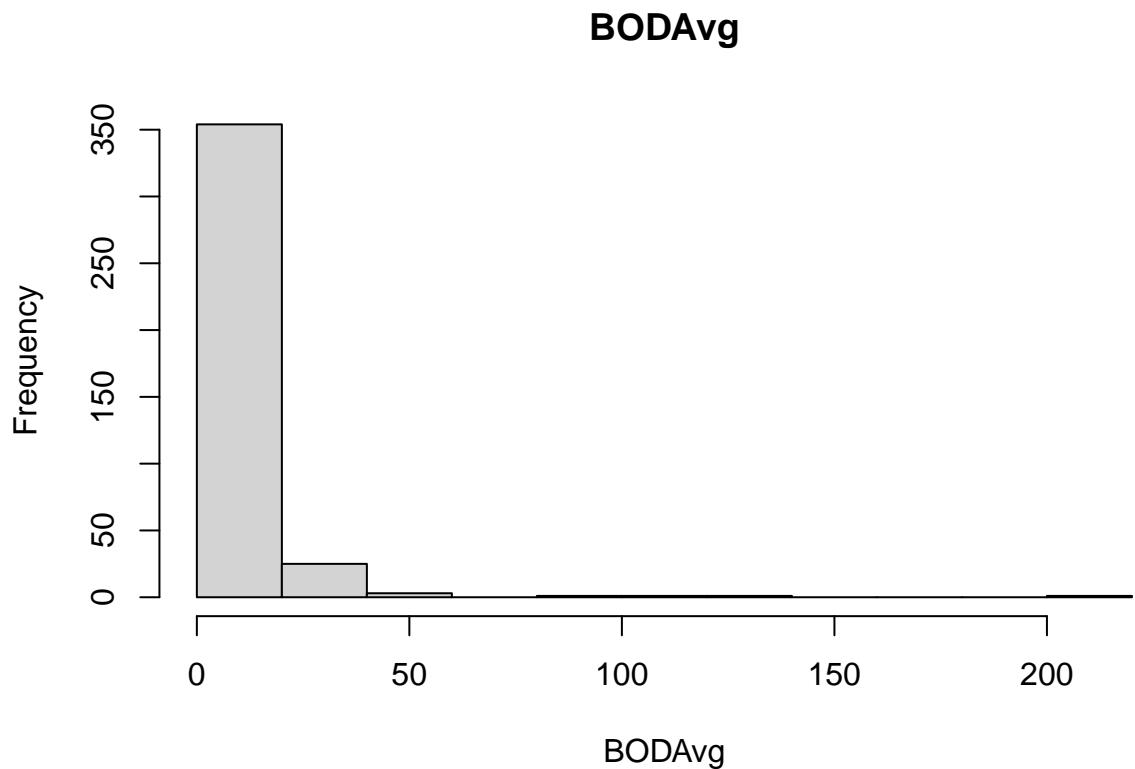
DissolvedO2Avg



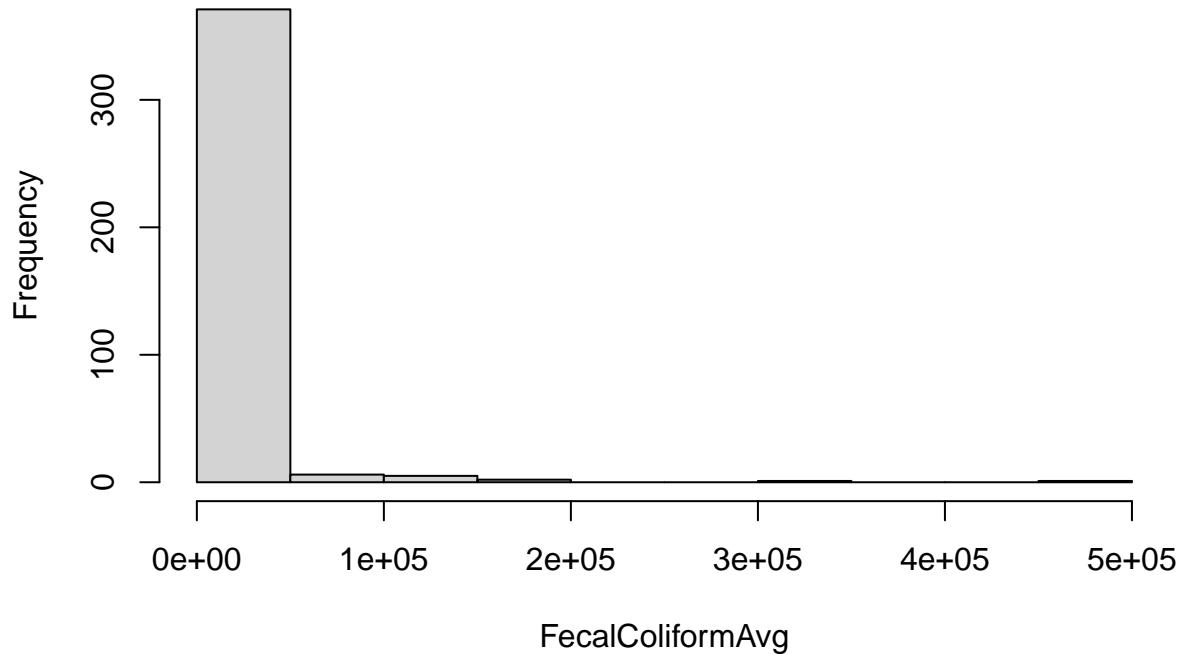


ConductivityAvg

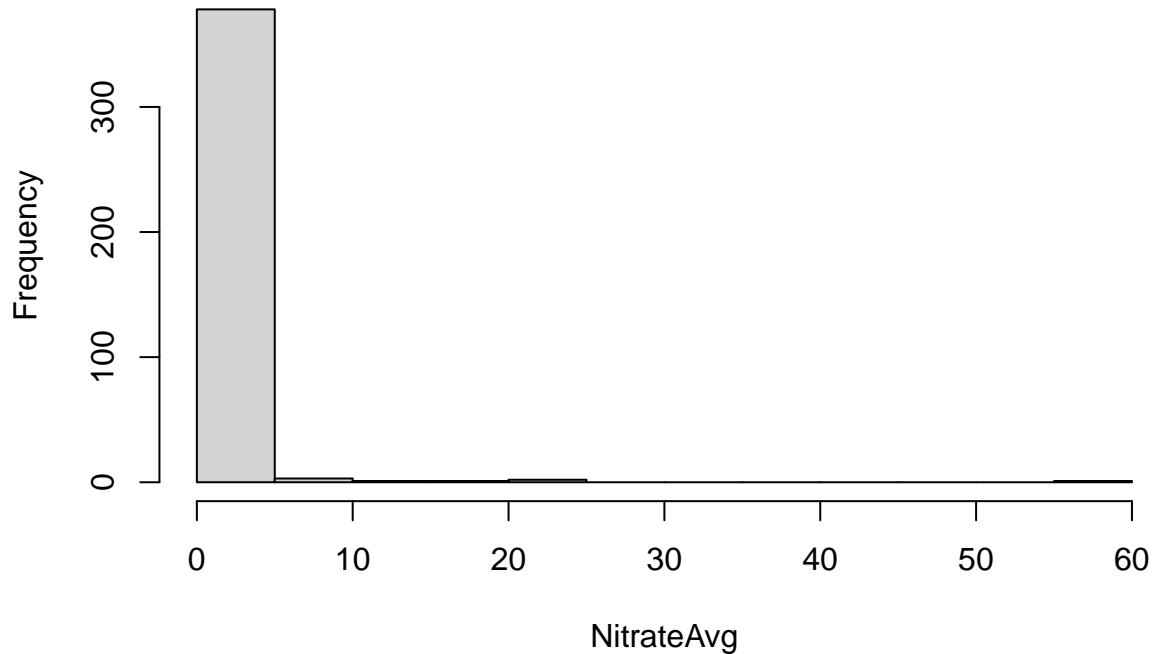




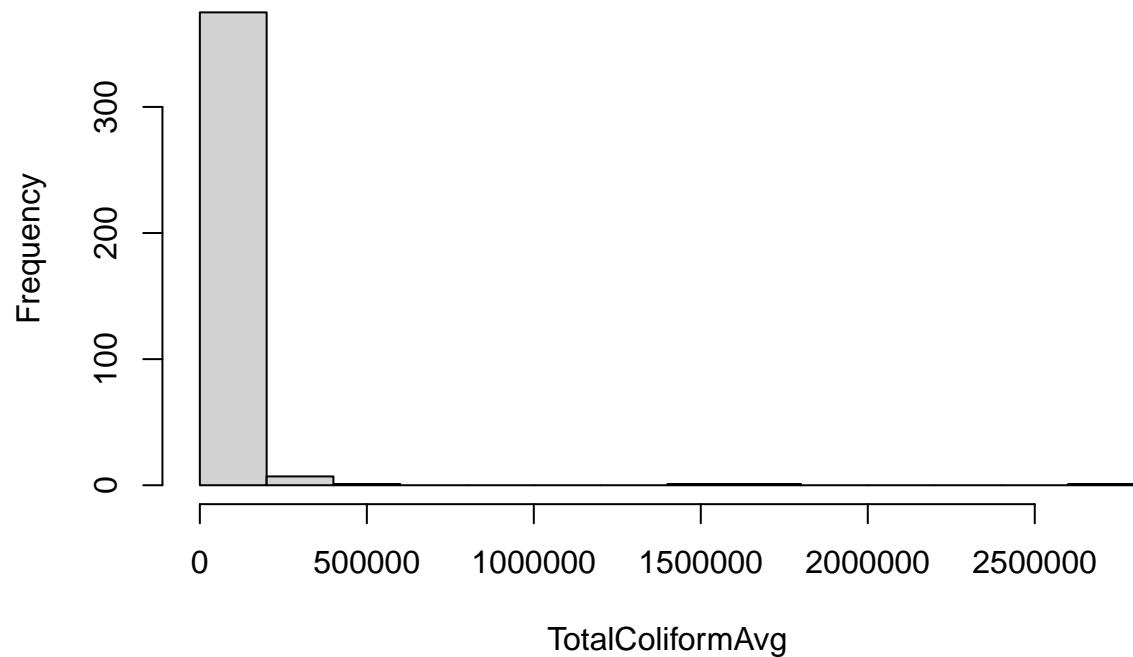
FecalColiformAvg

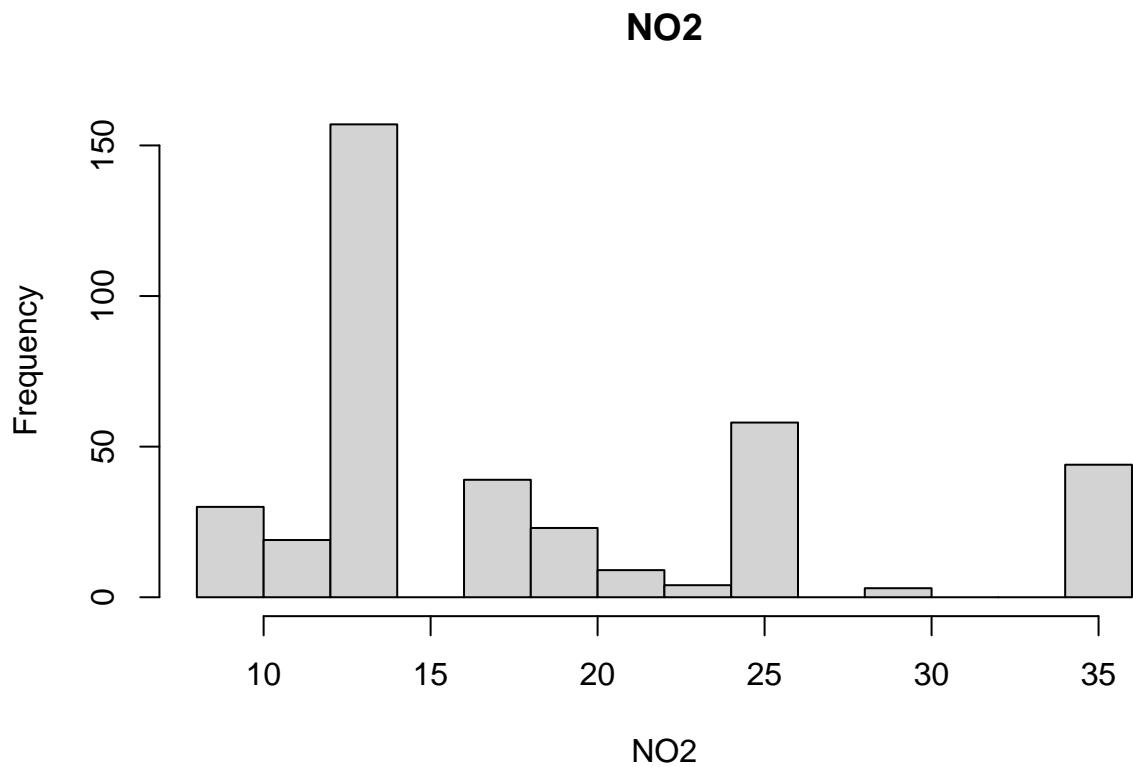


NitrateAvg

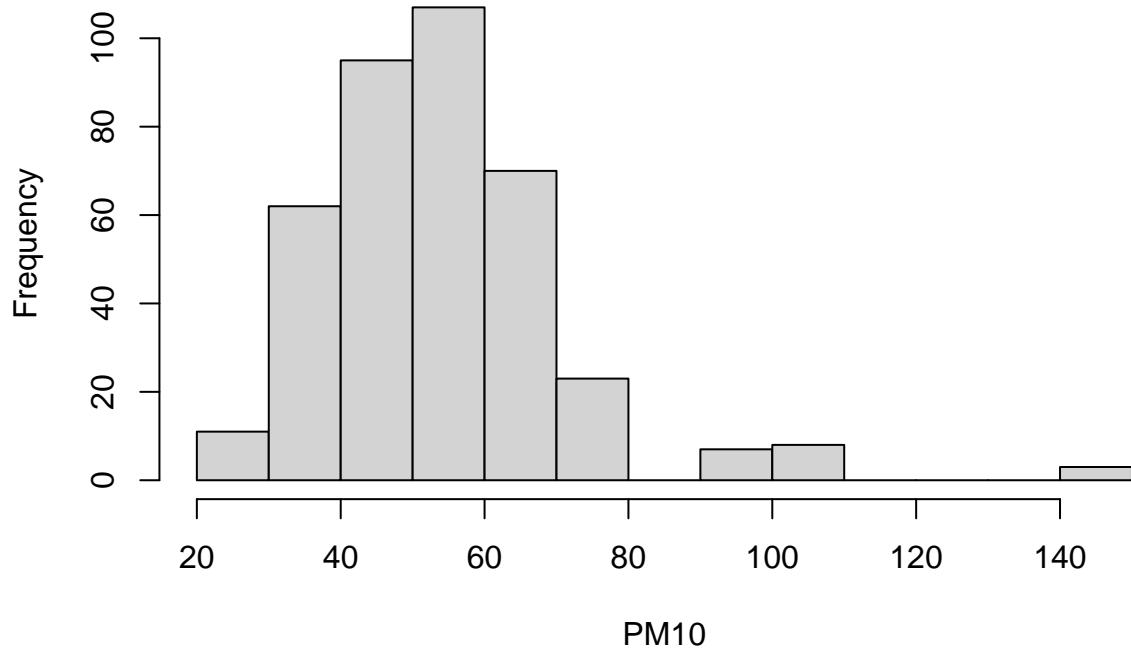


TotalColiformAvg

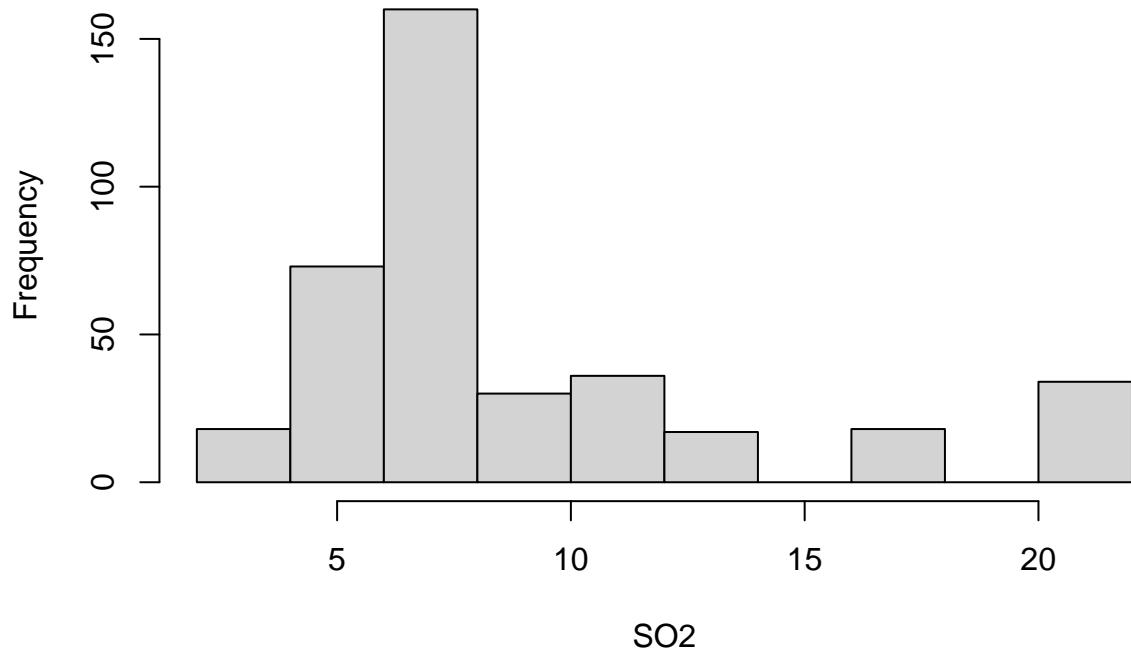




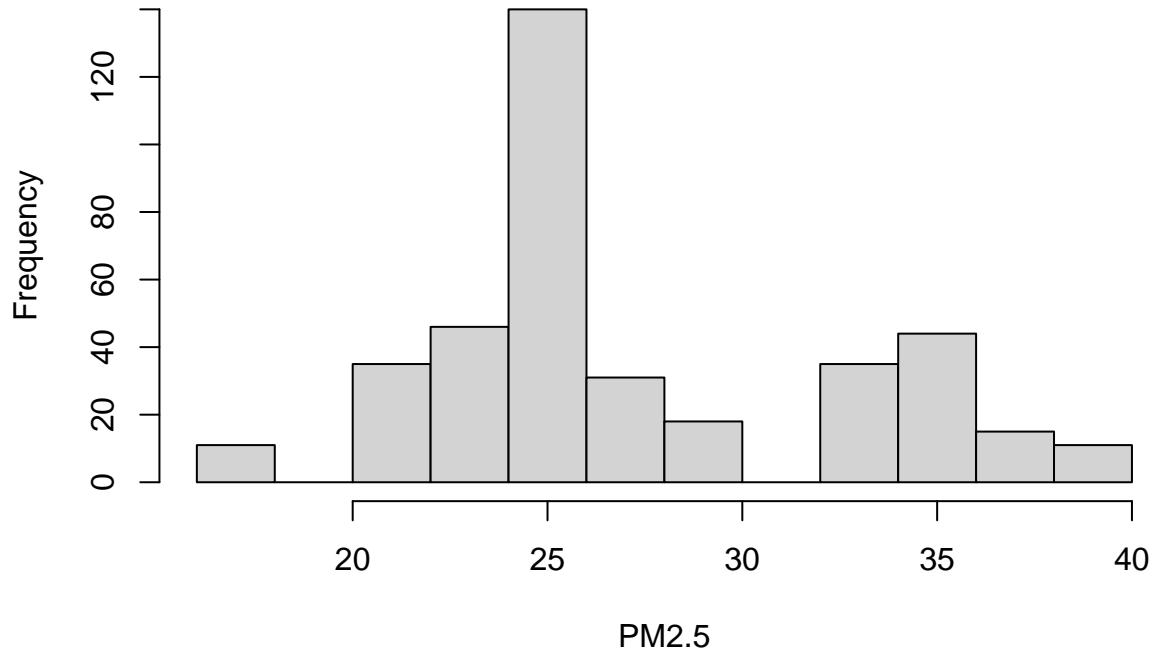
PM10



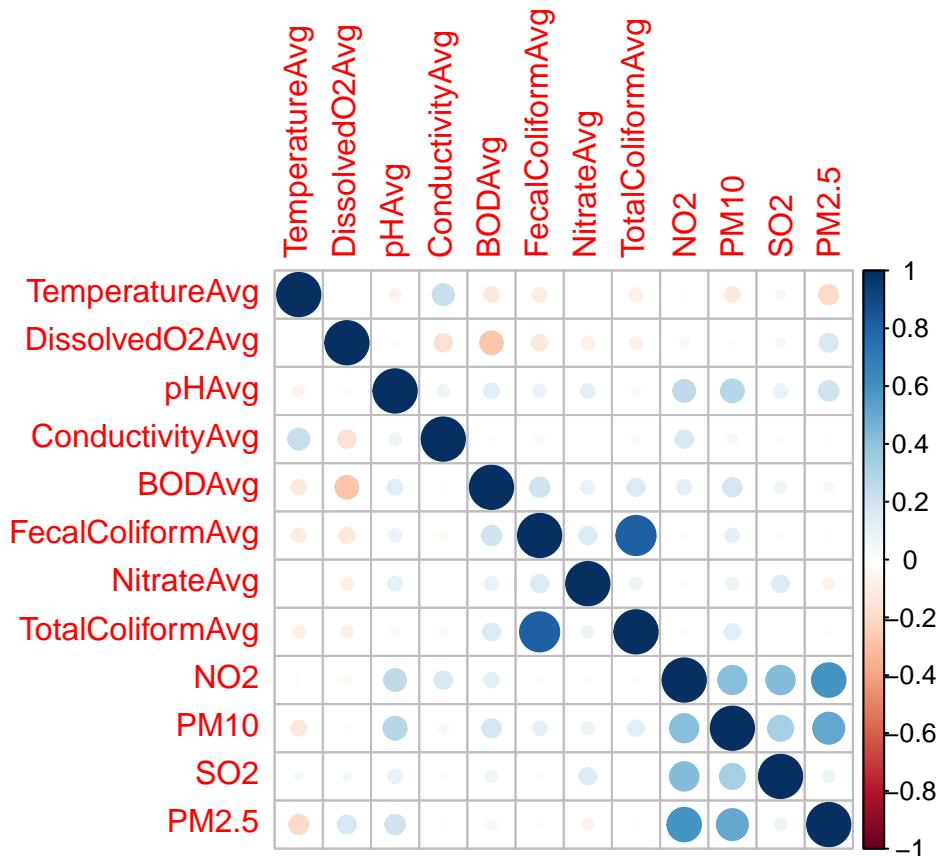
SO₂



PM2.5



```
cov_mat = cov(waterQuality_2019_Avg[, numCols])  
  
corr_matrix = cor(waterQuality_2019_Avg[, numCols])  
  
library(corrplot)  
  
corrplot(corr_matrix)
```



```
waterQuality_2019_Avg = aggregate(cbind(TemperatureAvg,
                                         DissolvedO2Avg,
                                         pHAvg,
                                         ConductivityAvg,
                                         BODAvg,
                                         NitrateAvg,
                                         TotalColiformAvg,
                                         FecalColiformAvg,
                                         NO2,
                                         PM10,
                                         SO2,
                                         PM2.5) ~ State,
                                     data = waterQuality_2019_Avg,
                                     mean)
```

```
pca_data = waterQuality_2019_Avg[, c("TemperatureAvg"
, "DissolvedO2Avg"
, "pHAvg"
, "ConductivityAvg"
, "BODAvg"
, "NitrateAvg"
, "FecalColiformAvg"
, "TotalColiformAvg"
, "NO2"
, "PM10"
, "SO2"]
```

```

,"PM2.5")]

# Center and scale the data
pca_data_scaled = scale(pca_data)

# Perform PCA
pca_result = princomp(pca_data_scaled, center = TRUE, scale. = TRUE)

## Warning: In princomp.default(pca_data_scaled, center = TRUE, scale. = TRUE) :
##   extra arguments 'center', 'scale.' will be disregarded

# Print the summary of PCA results
summary(pca_result)

## Importance of components:
##                               Comp.1    Comp.2    Comp.3    Comp.4    Comp.5
## Standard deviation      1.9591177 1.5337974 1.2415924 1.1499651 0.85667074
## Proportion of Variance 0.3358374 0.2058468 0.1348858 0.1157117 0.06421492
## Cumulative Proportion  0.3358374 0.5416842 0.6765700 0.7922817 0.85649661
##                               Comp.6    Comp.7    Comp.8    Comp.9    Comp.10
## Standard deviation      0.7757489 0.59234968 0.53782571 0.47204624 0.35988834
## Proportion of Variance 0.0526563 0.03070184 0.02530994 0.01949742 0.01133297
## Cumulative Proportion  0.9091529 0.93985475 0.96516469 0.98466211 0.99599508
##                               Comp.11   Comp.12
## Standard deviation      0.211363704 3.310426e-02
## Proportion of Variance 0.003909029 9.589053e-05
## Cumulative Proportion  0.999904109 1.000000e+00

#reverse the signs
pca_result$rotation = -1*pca_result$rotation

#display principal components
pca_result$rotation

## numeric(0)

var_explained = pca_result$sdev^2 / sum(pca_result$sdev^2)

print(var_explained)

##           Comp.1    Comp.2    Comp.3    Comp.4    Comp.5    Comp.6
## 3.358374e-01 2.058468e-01 1.348858e-01 1.157117e-01 6.421492e-02 5.265630e-02
##           Comp.7    Comp.8    Comp.9    Comp.10   Comp.11   Comp.12
## 3.070184e-02 2.530994e-02 1.949742e-02 1.133297e-02 3.909029e-03 9.589053e-05

pc.comp = pca_result$scores

#88%
pc.comp1 = -1*pc.comp[,1]
pc.comp2 = -1*pc.comp[,2]

```

```

pc.comp3 = -1*pc.comp[,3]
pc.comp4 = -1*pc.comp[,4]
pc.comp5 = -1*pc.comp[,5]
pc.comp6 = -1*pc.comp[,6]

library(factoextra)
library(tidyr)
library(ggplot2)

X = cbind(pc.comp1, pc.comp2, pc.comp3, pc.comp4, pc.comp5, pc.comp6)

##          pc.comp1    pc.comp2    pc.comp3    pc.comp4    pc.comp5    pc.comp6
## [1,] -0.42338677 -2.0254335 -0.57491156 -0.68866706  0.03373105  1.43162119
## [2,] -0.37888806 -0.4746289  1.24046204  0.42583132  0.18335981  0.32185762
## [3,] -0.34830087  1.8838367  0.08043478 -0.24484545  0.70502055 -0.70459974
## [4,] -0.21798982 -1.3713559 -1.80033482 -1.53654833  0.62198003  0.34688459
## [5,] -0.49460567 -2.2683292 -0.67350613 -0.94766986 -0.72837022 -0.34574849
## [6,] -0.13382215 -0.1608466 -2.12270416 -0.02504963  0.25803396  0.02736947
## [7,]  8.32240889 -0.2690643  0.32592711 -0.57922610 -0.66375870 -0.34073053
## [8,] -0.83514275  1.7277293  2.24655363 -1.29845216 -0.70457648 -0.03909843
## [9,] -0.79274782  3.6194762  0.50428580 -1.96831427 -0.83958331  0.50778820
## [10,] -0.01582507  2.4515910 -1.59968553  2.45858596 -0.34923191 -0.65323218
## [11,] -1.18225105 -0.7038423  0.42648292 -0.04523364 -0.40429086  0.25452729
## [12,] -1.08095855 -1.7533223  0.72723998 -0.20698773  0.16173433 -0.71202859
## [13,] -0.67864363 -0.3550360 -0.20032171  0.67392268 -0.09655650  1.60823225
## [14,]  0.03427609  0.6324116 -2.29854471  0.25074422  0.53846200 -0.82234761
## [15,] -1.15968825  0.4783996  0.62491103  0.61244026  0.59162588 -0.56796810
## [16,] -0.44031845 -0.9899853  2.40470688  1.00929472  0.21299758 -0.31398245
## [17,] -1.11213660 -0.5464175 -0.32296986 -1.94962905 -0.14655997 -1.06411691
## [18,]  1.59862214 -0.3123051  1.23966726  0.87277496  2.44775668  0.53628133
## [19,]  0.34047593  2.2236797 -0.51635803  0.13773215  0.62845047  1.12655113
## [20,] -0.87605482 -1.0740455  0.43837398  0.94644462 -0.26648514 -1.25311543
## [21,] -0.12502271 -0.7125117 -0.14970890  2.10285240 -2.18373925  0.65585541

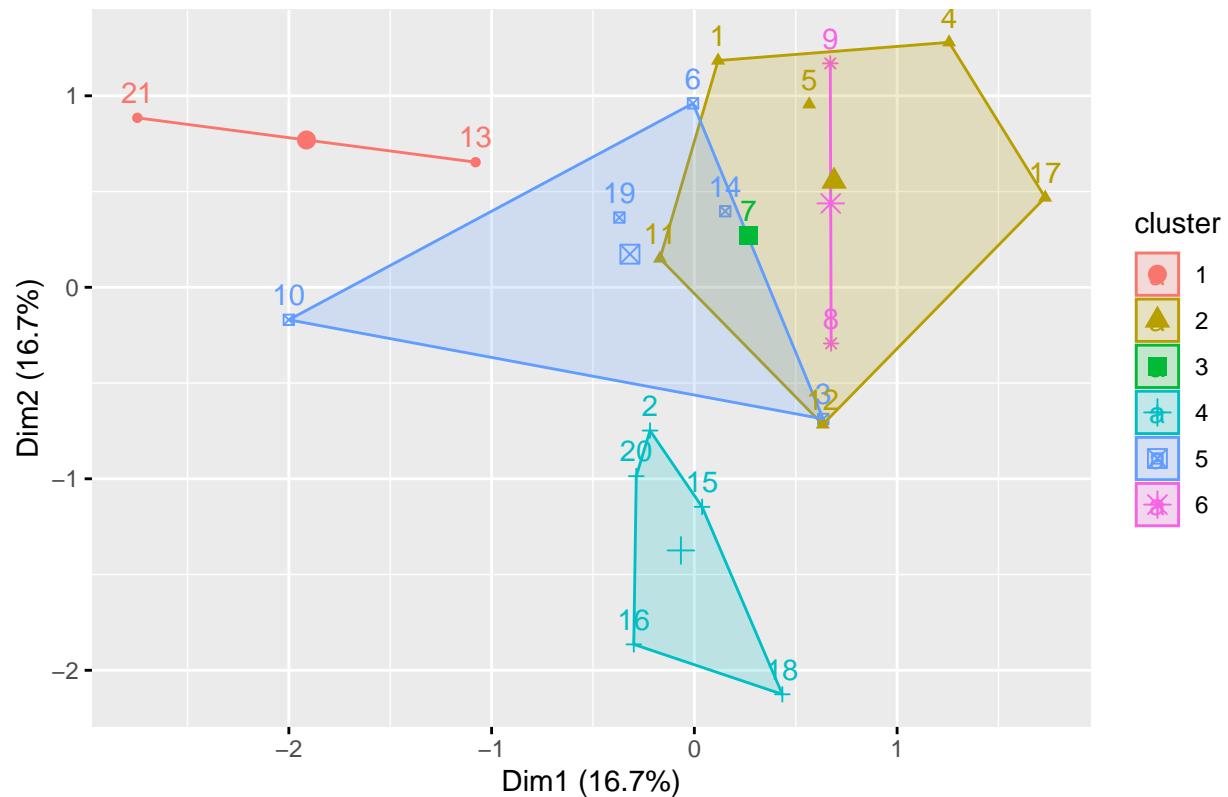
k = 6
kmeans_model = kmeans(X, centers = k, nstart=1)

waterQuality_2019_Avg$cluster=kmeans_model$cluster

fviz_cluster(kmeans_model, data = X)

```

Cluster plot



```
mean_by_group = aggregate(cbind(TemperatureAvg, DissolvedO2Avg, pHAvg, ConductivityAvg, BODAvg, NitrateAvg),
                          by = list(waterQuality_2019_Avg$State), FUN = mean)
table(waterQuality_2019_Avg$State, waterQuality_2019_Avg$cluster)
```

	1	2	3	4	5	6
ANDHRA PRADESH	0	1	0	0	0	0
ASSAM	0	0	0	1	0	0
BIHAR	0	0	0	0	1	0
DAMAN AND DIU, \r\nDADRA AND\r\nNAGAR HAVELI	0	1	0	0	0	0
GOA	0	1	0	0	0	0
GUJARAT	0	0	0	0	1	0
HARYANA	0	0	1	0	0	0
HIMACHAL\r\nPRADESH	0	0	0	0	0	1
JAMMU &\r\nKASHMIR	0	0	0	0	0	1
JHARKHAND	0	0	0	0	1	0
KARNATAKA	0	1	0	0	0	0
KERALA	0	1	0	0	0	0
MADHYA PRADESH	1	0	0	0	0	0
MAHARASHTRA	0	0	0	1	0	0
MANIPUR	0	0	0	1	0	0
MEGHALAYA	0	0	0	1	0	0
PUDUCHERRY	0	1	0	0	0	0
PUNJAB	0	0	0	1	0	0
RAJASTHAN	0	0	0	0	1	0

```

##    TAMIL NADU          0 0 0 1 0 0
##    TRIPURA            1 0 0 0 0 0

```

waterQuality_2019_Avg

		State	TemperatureAvg	DissolvedO2Avg	
##	## 1	ANDHRA PRADESH	23.09091	6.513636	
##	## 2	ASSAM	18.37500	6.076481	
##	## 3	BIHAR	19.25000	6.025000	
##	## 4	DAMAN AND DIU, \r\nDADRA AND\r\nNAGAR HAVELI	27.94444	5.627778	
##	## 5	GOA	27.03846	6.627379	
##	## 6	GUJARAT	25.11111	6.273393	
##	## 7	HARYANA	19.53381	4.565461	
##	## 8	HIMACHAL\r\nPRADESH	13.15694	8.060000	
##	## 9	JAMMU &\r\nKASHMIR	19.66667	8.166667	
##	## 10	JHARKHAND	14.31250	6.795231	
##	## 11	KARNATAKA	25.00000	6.988889	
##	## 12	KERALA	24.36986	6.021918	
##	## 13	MADHYA PRADESH	21.24444	6.772222	
##	## 14	MAHARASHTRA	22.05556	6.221491	
##	## 15	MANIPUR	17.18750	6.806250	
##	## 16	MEGHALAYA	13.31667	6.276667	
##	## 17	PUDUCHERRY	27.25000	6.625000	
##	## 18	PUNJAB	13.11588	4.181086	
##	## 19	RAJASTHAN	16.83333	6.311846	
##	## 20	TAMIL NADU	24.50000	6.200000	
##	## 21	TRIPURA	21.00000	6.975000	
	pHAvg	ConductivityAvg	BODAvg	NitrateAvg	TotalColiformAvg
## 1	7.722727	10041.3182	2.386364	1.8500000	430.09091
## 2	7.325000	183.4531	5.896875	1.5500000	7532.58286
## 3	7.650000	484.3750	2.637500	0.1850000	49825.00000
## 4	7.838889	12905.2778	6.438789	1.1500000	363.61111
## 5	7.044231	12424.5000	6.956138	1.6460675	13856.75000
## 6	7.655556	9185.3889	14.330556	0.3913889	450.86111
## 7	7.850000	4463.2500	60.656250	5.8456515	723425.00000
## 8	7.700000	955.2000	1.103333	0.6720424	106913.13333
## 9	8.150000	141.0000	4.704092	0.8134746	28351.32574
## 10	7.193750	2555.8704	2.968750	1.2152119	14691.28787
## 11	7.377778	213.8889	3.738889	0.5527778	716.94444
## 12	6.959589	4114.8630	1.734247	0.6756849	6551.45890
## 13	7.850000	544.1111	1.644444	0.7766667	83.16667
## 14	7.420833	9562.8611	6.719444	1.6025177	2167.48611
## 15	7.278125	373.2656	3.468750	0.1050132	233.26562
## 16	6.900000	188.1000	8.253333	0.7841667	38921.58333
## 17	7.575000	8045.0000	5.242103	0.3225000	28351.32574
## 18	7.571875	890.3125	30.125000	1.1868750	42156.25000
## 19	8.116667	616.6667	6.956138	1.8800000	75.33333
## 20	6.675000	438.1429	5.872241	0.9825000	73.07143
## 21	7.111538	146.0962	1.598077	3.2169231	209.23077
	FecalColiformAvg	NO2	PM10	S02	PM2.5 cluster
## 1	653.05743	17.562500	28.25000	7.125000	17.56250 2
## 2	1346.43750	13.429487	58.08731	7.089744	25.67907 4
## 3	12306.25000	22.136364	94.04545	8.818922	39.12783 5
## 4	279.05556	18.052259	55.41188	8.969022	27.37201 2

```

## 5      8883.42814 13.411765 59.06424 7.021524 21.26011      2
## 6      653.14621 24.562963 67.99259 16.120823 29.08395      5
## 7     126268.75000 24.576923 78.03846 12.346154 26.32933      3
## 8     23406.66667 11.119048 71.55952 2.226190 36.89954      6
## 9      7069.63172 18.000000 148.66667 4.000000 39.28123      6
## 10     1851.78293 34.255556 104.85260 21.177778 35.46457      5
## 11     195.22222 17.102624 47.58049 7.824026 24.67611      2
## 12     3381.01370 12.891386 43.55600 4.900582 24.99776      2
## 13     13.33333 20.947368 40.36842 13.243725 20.94737      1
## 14     211.97222 34.694938 67.94619 11.944126 35.15770      5
## 15     252.44162 24.231229 38.50000 7.254072 32.64061      4
## 16     16453.51667 9.928571 39.37500 6.562912 23.85490      4
## 17     7069.63172 11.500000 41.83333 6.500000 39.28123      2
## 18     11884.37500 17.832208 67.48132 7.546342 23.52170      4
## 19     44.00000 28.032680 99.91756 9.937908 32.46874      5
## 20     532.40227 18.052259 55.41188 8.969022 27.37201      4
## 21     109.73077 13.000000 55.75000 21.000000 24.25000      1

```

```

Result_2019<- mean_by_group[order(mean_by_group$TotalColiformAvg, mean_by_group$BODAvg, mean_by_group$Di

Result_2019$Rank=c("A", "B", "C", "D", "E", "F")

class(Result_2019$cluster)

## [1] "integer"

class(waterQuality_2019_Avg$cluster)

## [1] "integer"

WaterAir2019 = merge(waterQuality_2019_Avg, Result_2019[,c("cluster","Rank")], by = "cluster", all.x = T

WaterQualityAlone = waterQuality_2019_Avg[, c("TemperatureAvg"
,"DissolvedO2Avg"
,"pHAvg"
,"ConductivityAvg"
,"BODAvg"
,"FecalColiformAvg"
,"NitrateAvg"
,"TotalColiformAvg"
)]

# Center and scale the data
WaterQualityAlone_scaled = scale(WaterQualityAlone)

# Perform PCA
pca_result = princomp(WaterQualityAlone_scaled, center = TRUE, scale. = TRUE)

## Warning: In princomp.default(WaterQualityAlone_scaled, center = TRUE, scale. = TRUE) :
##   extra arguments 'center', 'scale.' will be disregarded

```

```

# Print the summary of PCA results
summary(pca_result)

## Importance of components:
##                               Comp.1     Comp.2     Comp.3     Comp.4     Comp.5
## Standard deviation      1.9314039 1.2660204 0.9991275 0.8088802 0.54596668
## Proportion of Variance 0.4896046 0.2103685 0.1310211 0.0858752 0.03912295
## Cumulative Proportion  0.4896046 0.6999731 0.8309942 0.9168694 0.95599235
##                               Comp.6     Comp.7     Comp.8
## Standard deviation      0.52287086 0.243055773 0.0531629572
## Proportion of Variance 0.03588295 0.007753739 0.0003709519
## Cumulative Proportion  0.99187531 0.999629048 1.000000000000

#reverse the signs
pca_result$rotation = -1*pca_result$rotation

#display principal components
pca_result$rotation

## numeric(0)

var_explained = pca_result$sdev^2 / sum(pca_result$sdev^2)

print(var_explained)

##                               Comp.1     Comp.2     Comp.3     Comp.4     Comp.5     Comp.6
## 0.4896046178 0.2103685117 0.1310210803 0.0858751954 0.0391229493 0.0358829543
##                               Comp.7     Comp.8
## 0.0077537393 0.0003709519

pc.comp = pca_result$scores

#86%
pc.comp1 = -1*pc.comp[,1]
pc.comp2 = -1*pc.comp[,2]
pc.comp3 = -1*pc.comp[,3]
pc.comp4 = -1*pc.comp[,4]
pc.comp5= -1*pc.comp[,5]
pc.comp6 = -1*pc.comp[,6]

library(factoextra)
library(tidyr)
library(ggplot2)

X = cbind(pc.comp1, pc.comp2,pc.comp3, pc.comp4,pc.comp5,pc.comp6)

X

##          pc.comp1    pc.comp2    pc.comp3    pc.comp4    pc.comp5    pc.comp6
## [1,] -0.34728680 -1.34440233 -0.746219355  0.05835626  0.81772745 -0.20617820
## [2,] -0.23833725  0.77806745  0.602221042  0.20380588  0.08273259 -0.52722422

```

```

## [3,] -0.38408005  0.69535336 -0.176143796  0.67910553 -0.65856585  0.19116206
## [4,] -0.16817015 -2.62096089 -0.644227553  0.89215648  0.13338708 -0.08649049
## [5,] -0.37335127 -2.24751562  0.622741193 -0.86382861  0.54465596  0.38123481
## [6,] -0.38599110 -1.50357715 -0.425181030  0.60665793 -0.24799552  0.45321485
## [7,]  8.27905361 -0.14137056 -0.217875137 -0.76360684 -0.27470566  0.08706358
## [8,] -0.48311166  1.88937315 -1.316522857 -0.85623520  0.42701263  0.99556716
## [9,] -0.84392272  1.04694015 -2.283833009 -0.56606728 -0.36281708 -0.05268209
## [10,] -0.63312135  1.14403389  0.484222036 -0.19518746  0.86599264  0.34219530
## [11,] -1.10200837  0.03292164 -0.007927365 -0.55060409 -1.02166839 -0.27885714
## [12,] -0.82656392 -0.65937084  1.285415761 -0.22122216 -0.44330243  0.07580659
## [13,] -0.81745709  0.48148115 -0.943163037  0.21336838 -0.49919327 -0.41895092
## [14,] -0.25757361 -1.17738510  0.099152089  0.15675990  0.77344026  0.06011281
## [15,] -1.09926888  1.11099797  0.329715777  0.03072848 -0.18452860  0.47677292
## [16,] -0.16739889  1.59739337  1.425584053 -0.05577068  0.36983753  0.70668287
## [17,] -0.72616325 -1.53527188 -0.477006451 -0.08982932 -0.60512831  0.40627864
## [18,]  1.62709870  1.08876628  1.071241360  2.49495318  0.10177291 -0.06117025
## [19,]  0.06375282  0.96955388 -1.250357839  0.85589813  0.34494006 -0.90269613
## [20,] -0.81851282 -0.10756720  1.943317942 -0.70009862 -0.74045197 -0.27935766
## [21,] -0.29758594  0.50253930  0.624846179 -1.32933988  0.57685797 -1.36248448

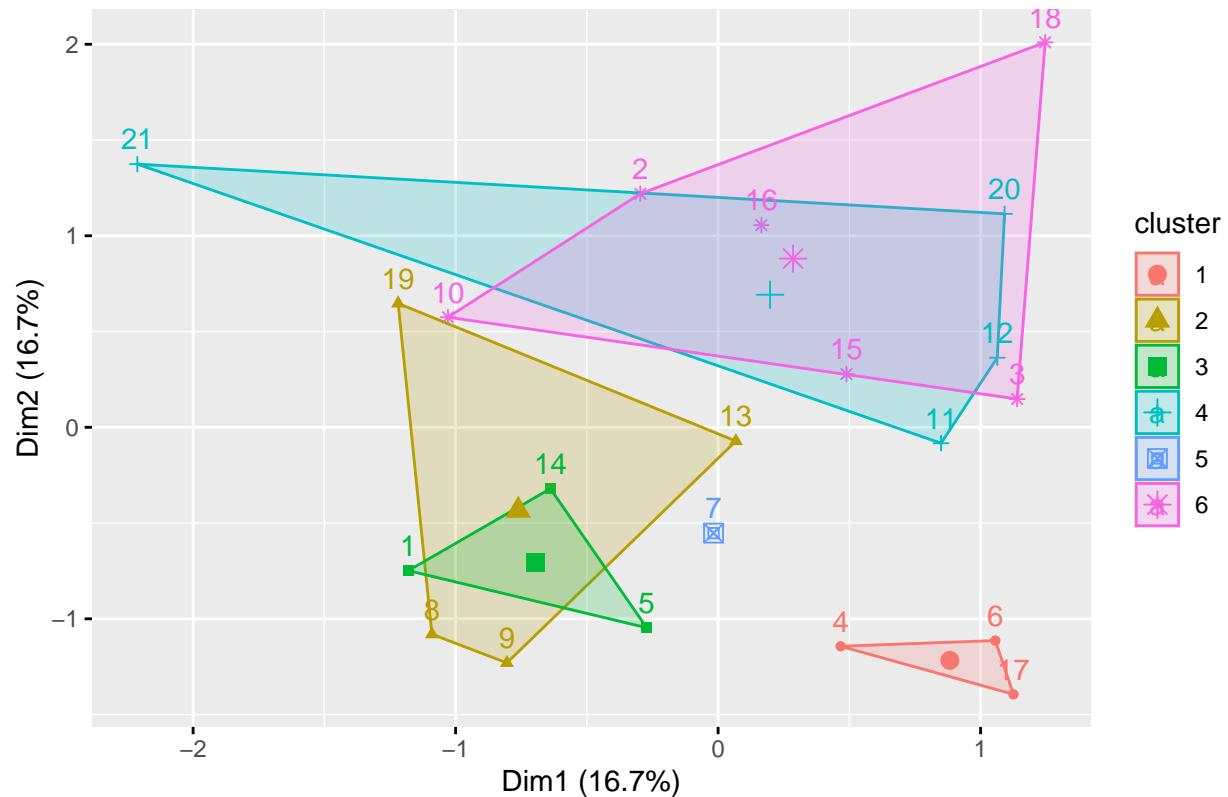
k =6
kmeans_model = kmeans(X, centers = k,nstart=1)

waterQuality_2019_Avg$cluster=kmeans_model$cluster

fviz_cluster(kmeans_model, data = X)

```

Cluster plot



```
mean_by_group = aggregate(cbind(TemperatureAvg, DissolvedO2Avg, pHAvg, ConductivityAvg, BODAvg, NitrateAvg),
                          by = list(waterQuality_2019_Avg$State), FUN = mean)
table(waterQuality_2019_Avg$State, waterQuality_2019_Avg$cluster)
```

```
##
##          1 2 3 4 5 6
## ANDHRA PRADESH 0 0 1 0 0 0
## ASSAM        0 0 0 0 0 1
## BIHAR         0 0 0 0 0 1
## DAMAN AND DIU, \r\nDADRA AND\r\nNAGAR HAVELI 1 0 0 0 0 0
## GOA          0 0 1 0 0 0
## GUJARAT      1 0 0 0 0 0
## HARYANA      0 0 0 0 1 0
## HIMACHAL\r\nPRADESH 0 1 0 0 0 0
## JAMMU &\r\nKASHMIR 0 1 0 0 0 0
## JHARKHAND    0 0 0 0 0 1
## KARNATAKA    0 0 0 1 0 0
## KERALA        0 0 0 1 0 0
## MADHYA PRADESH 0 1 0 0 0 0
## MAHARASHTRA  0 0 1 0 0 0
## MANIPUR       0 0 0 0 0 1
## MEGHALAYA    0 0 0 0 0 1
## PUDUCHERRY  1 0 0 0 0 0
## PUNJAB        0 0 0 0 0 1
## RAJASTHAN    0 1 0 0 0 0
```

```

##    TAMIL NADU          0 0 0 1 0 0
##    TRIPURA            0 0 0 1 0 0

```

WaterQualityAlone

```

##    TemperatureAvg DissolvedO2Avg      pHAvg ConductivityAvg      BODAvg
## 1      23.09091     6.513636 7.722727     10041.3182    2.386364
## 2      18.37500     6.076481 7.325000      183.4531    5.896875
## 3      19.25000     6.025000 7.650000      484.3750    2.637500
## 4      27.94444     5.627778 7.838889     12905.2778    6.438789
## 5      27.03846     6.627379 7.044231     12424.5000    6.956138
## 6      25.11111     6.273393 7.655556     9185.3889   14.330556
## 7      19.53381     4.565461 7.850000     4463.2500   60.656250
## 8      13.15694     8.060000 7.700000      955.2000   1.103333
## 9      19.66667     8.166667 8.150000     141.0000   4.704092
## 10     14.31250     6.795231 7.193750     2555.8704   2.968750
## 11     25.00000     6.988889 7.377778     213.8889   3.738889
## 12     24.36986     6.021918 6.959589     4114.8630   1.734247
## 13     21.24444     6.772222 7.850000      544.1111   1.644444
## 14     22.05556     6.221491 7.420833     9562.8611   6.719444
## 15     17.18750     6.806250 7.278125      373.2656   3.468750
## 16     13.31667     6.276667 6.900000      188.1000   8.253333
## 17     27.25000     6.625000 7.575000     8045.0000   5.242103
## 18     13.11588     4.181086 7.571875     890.3125 30.125000
## 19     16.83333     6.311846 8.116667      616.6667   6.956138
## 20     24.50000     6.200000 6.675000      438.1429   5.872241
## 21     21.00000     6.975000 7.111538     146.0962   1.598077
##    FecalColiformAvg NitrateAvg TotalColiformAvg
## 1      653.05743   1.8500000     430.09091
## 2      1346.43750   1.5500000     7532.58286
## 3      12306.25000   0.1850000    49825.00000
## 4      279.05556   1.1500000     363.61111
## 5      8883.42814   1.6460675    13856.75000
## 6      653.14621   0.3913889     450.86111
## 7      126268.75000  5.8456515   723425.00000
## 8      23406.66667   0.6720424   106913.13333
## 9      7069.63172   0.8134746   28351.32574
## 10     1851.78293   1.2152119   14691.28787
## 11     195.22222   0.5527778    716.94444
## 12     3381.01370   0.6756849   6551.45890
## 13     13.33333   0.7766667     83.16667
## 14     211.97222   1.6025177   2167.48611
## 15     252.44162   0.1050132   233.26562
## 16     16453.51667   0.7841667   38921.58333
## 17     7069.63172   0.3225000   28351.32574
## 18     11884.37500   1.1868750   42156.25000
## 19     44.00000   1.8800000     75.33333
## 20     532.40227   0.9825000     73.07143
## 21     109.73077   3.2169231   209.23077

```

```

Result_2019_water= mean_by_group[order(mean_by_group$TotalColiformAvg, mean_by_group$BODAvg, mean_by_gr

```

```

Result_2019_water$Rank=c("A", "B", "C", "D", "E", "F")
class(waterQuality_2019_Avg$cluster)

## [1] "integer"

class(waterQuality_2019_Avg$cluster)

## [1] "integer"

Water2019 = merge(waterQuality_2019_Avg, Result_2019_water[,c("cluster","Rank")], by = "cluster", all.x=TRUE)
waterQuality_2020=finalDataset[finalDataset$year == 2020,]
waterQuality_2020$TemperatureMin      =as.numeric(waterQuality_2020$TemperatureMin)

## Warning: NAs introduced by coercion

waterQuality_2020$TemperatureMax      =as.numeric(waterQuality_2020$TemperatureMin)
waterQuality_2020$DissolvedO2Min     =as.numeric(waterQuality_2020$DissolvedO2Min)

## Warning: NAs introduced by coercion

waterQuality_2020$DissolvedO2Max     =as.numeric(waterQuality_2020$DissolvedO2Max)

## Warning: NAs introduced by coercion

waterQuality_2020$pHMin              =as.numeric(waterQuality_2020$pHMin)

## Warning: NAs introduced by coercion

waterQuality_2020$pHMax              =as.numeric(waterQuality_2020$pHMax)

## Warning: NAs introduced by coercion

waterQuality_2020$ConductivityMin    =as.numeric(waterQuality_2020$ConductivityMin)

## Warning: NAs introduced by coercion

waterQuality_2020$ConductivityMax    =as.numeric(waterQuality_2020$ConductivityMax)

## Warning: NAs introduced by coercion

waterQuality_2020$BODMin             =as.numeric(waterQuality_2020$BODMin)

## Warning: NAs introduced by coercion

```

```

waterQuality_2020$BODMax           =as.numeric(waterQuality_2020$BODMax)

## Warning: NAs introduced by coercion

waterQuality_2020$NitrateMin       =as.numeric(waterQuality_2020$NitrateMin)

## Warning: NAs introduced by coercion

waterQuality_2020$NitrateMax       =as.numeric(waterQuality_2020$NitrateMax)

## Warning: NAs introduced by coercion

waterQuality_2020$FecalColiformMin =as.numeric(waterQuality_2020$FecalColiformMin)

## Warning: NAs introduced by coercion

waterQuality_2020$FecalColiformMax =as.numeric(waterQuality_2020$FecalColiformMax)

## Warning: NAs introduced by coercion

waterQuality_2020$TotalColiformMin =as.numeric(waterQuality_2020$TotalColiformMin)

## Warning: NAs introduced by coercion

waterQuality_2020$TotalColiformMax =as.numeric(waterQuality_2020$TotalColiformMax)

## Warning: NAs introduced by coercion

waterQuality_2020$NO2              =as.numeric(waterQuality_2020$NO2)
waterQuality_2020$PM10             =as.numeric(waterQuality_2020$PM10)

cols=c("TemperatureMin"
,"TemperatureMax"
,"DissolvedO2Min"
,"DissolvedO2Max"
,"pHMin"
,"pHMax"
,"ConductivityMin"
,"ConductivityMax"
,"BODMin"
,"BODMax"
,"NitrateMin"
,"NitrateMax"
,"FecalColiformMin"
,"FecalColiformMax"
,"TotalColiformMin"
,"TotalColiformMax"
,"NO2"
,
```

```

,"PM10"
,"SO2"
,"PM2.5")

waterQuality_2020_Avg=waterQuality_2020

# Calculate the average values for each variable
waterQuality_2020_Avg$TemperatureAvg = (waterQuality_2020_Avg$TemperatureMin + waterQuality_2020_Avg$TemperatureMax) / 2
waterQuality_2020_Avg$DissolvedO2Avg = (waterQuality_2020_Avg$DissolvedO2Min + waterQuality_2020_Avg$DissolvedO2Max) / 2
waterQuality_2020_Avg$pHAvg = (waterQuality_2020_Avg$pHMin + waterQuality_2020_Avg$pHMax) / 2
waterQuality_2020_Avg$ConductivityAvg = (waterQuality_2020_Avg$ConductivityMin + waterQuality_2020_Avg$ConductivityMax) / 2
waterQuality_2020_Avg$FecalColiformAvg = (waterQuality_2020_Avg$FecalColiformMin + waterQuality_2020_Avg$FecalColiformMax) / 2
waterQuality_2020_Avg$BODAvg = (waterQuality_2020_Avg$BODMin + waterQuality_2020_Avg$BODMax) / 2
waterQuality_2020_Avg$NitrateAvg = (waterQuality_2020_Avg$NitrateMin + waterQuality_2020_Avg$NitrateMax) / 2

waterQuality_2020_Avg$TotalColiformAvg = (waterQuality_2020_Avg$TotalColiformMin + waterQuality_2020_Avg$TotalColiformMax) / 2

waterQuality_2020_Avg = subset(waterQuality_2020_Avg, select=-c(TemperatureMin
, TemperatureMax
, DissolvedO2Min
, DissolvedO2Max
, pHMin
, pHMax
, ConductivityMin
, ConductivityMax
, BODMin
, BODMax
, NitrateMin
, NitrateMax
, FecalColiformMin
, FecalColiformMax
, TotalColiformMin
, TotalColiformMax))

numCols=c("TemperatureAvg"
,"DissolvedO2Avg"
,"pHAvg"
,"ConductivityAvg"
,"BODAvg"
,"FecalColiformAvg"
,"NitrateAvg"
,"TotalColiformAvg"
,"NO2"
,"PM10"
,"SO2"
,"PM2.5")

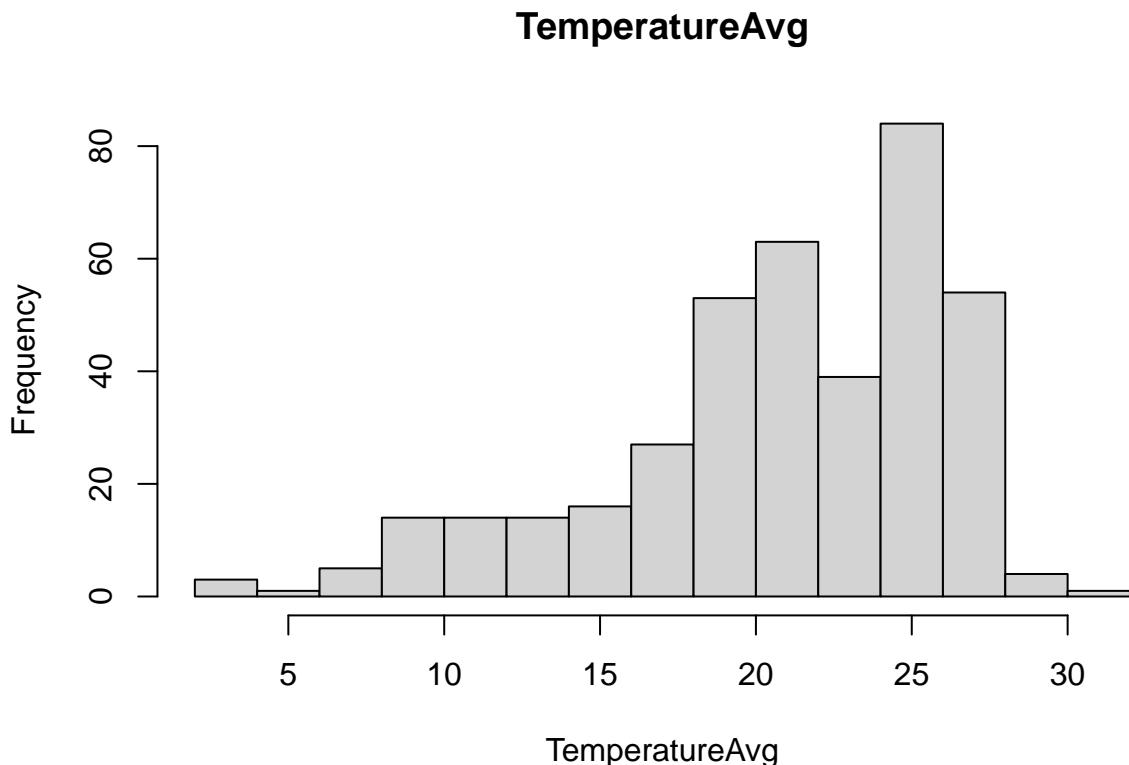
for (col in colnames(waterQuality_2020_Avg)) {
  # Check if column has any NAs
  if (sum(is.na(waterQuality_2020_Avg[, col])) > 0) {
    # Replace NAs with mean value of the column
    waterQuality_2020_Avg[is.na(waterQuality_2020_Avg[, col]), col] = mean(waterQuality_2020_Avg[, col])
  }
}

```

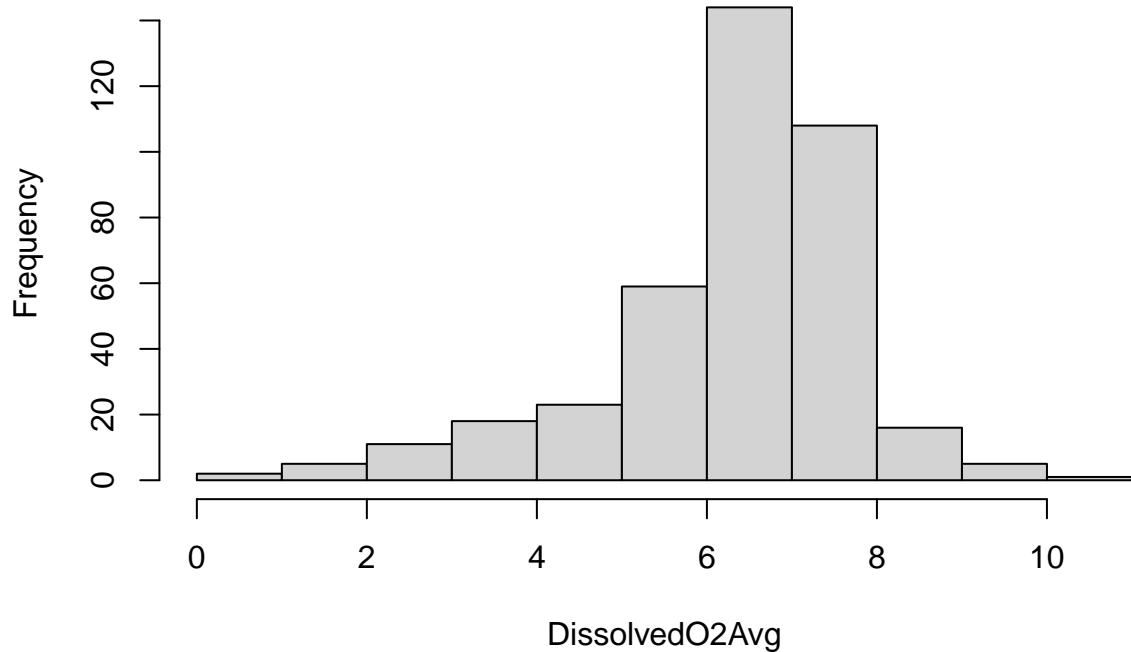
```
    }

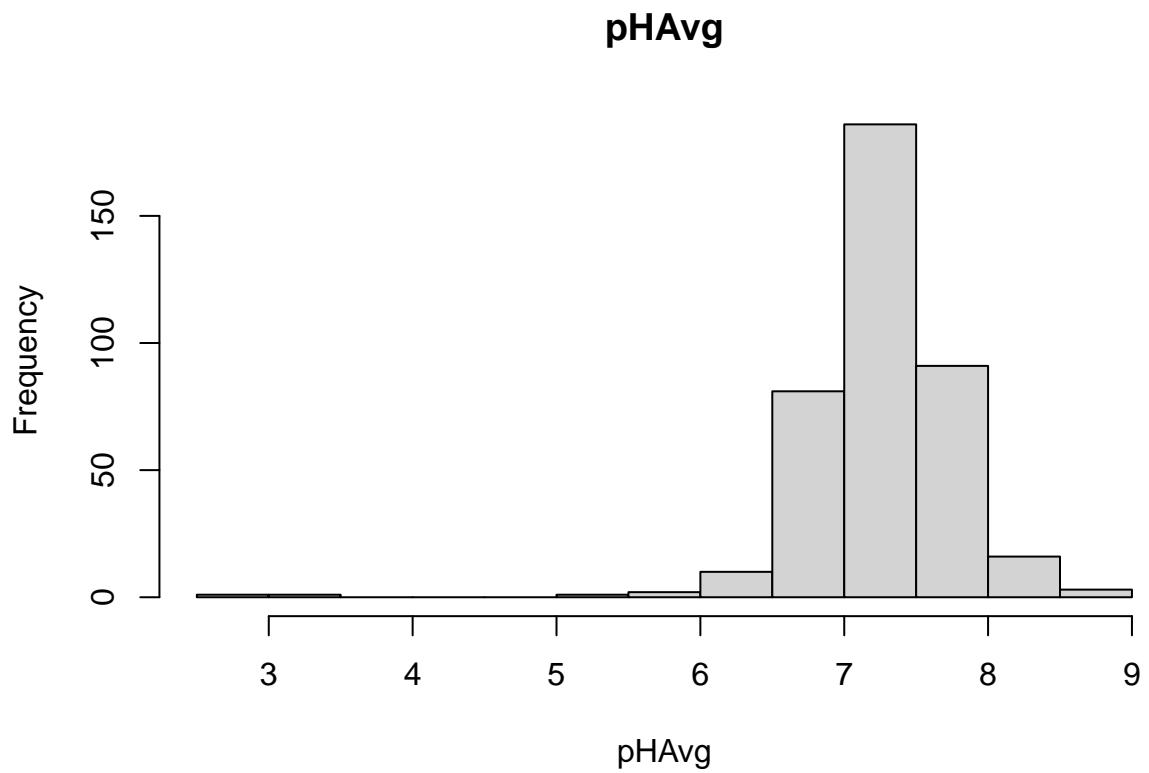
}

for (col in numCols) {
  # Create the histogram plot
  hist(waterQuality_2020_Avg[[col]], main = col, xlab = col)
}
```

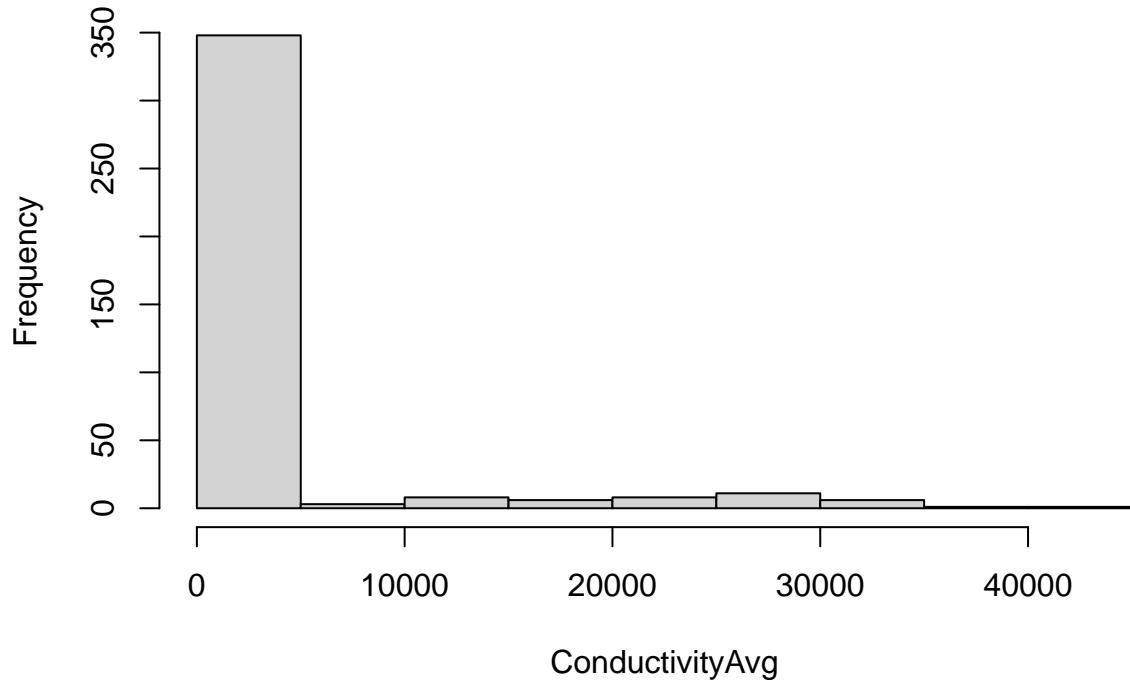


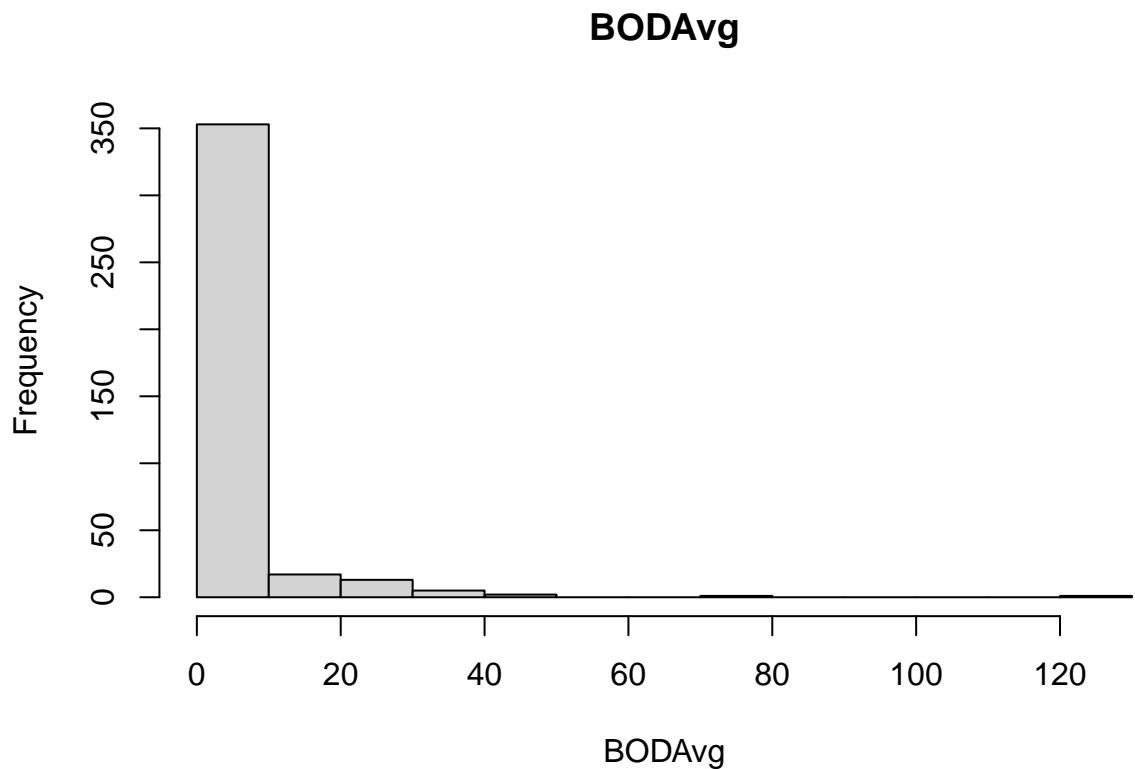
DissolvedO2Avg



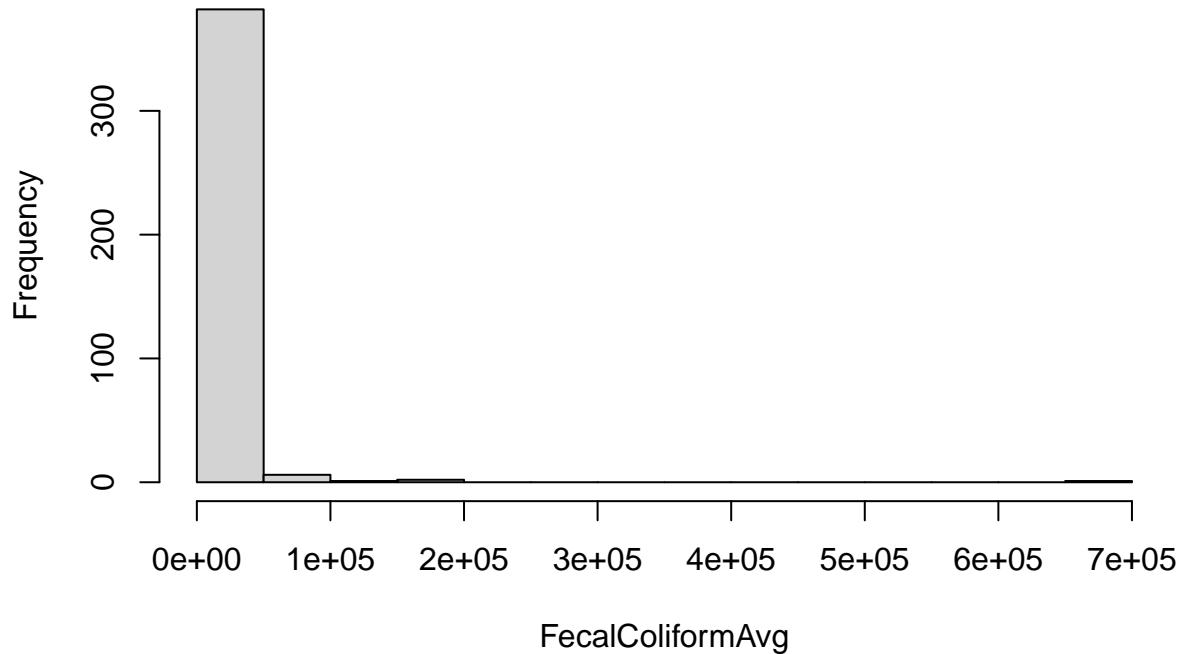


ConductivityAvg

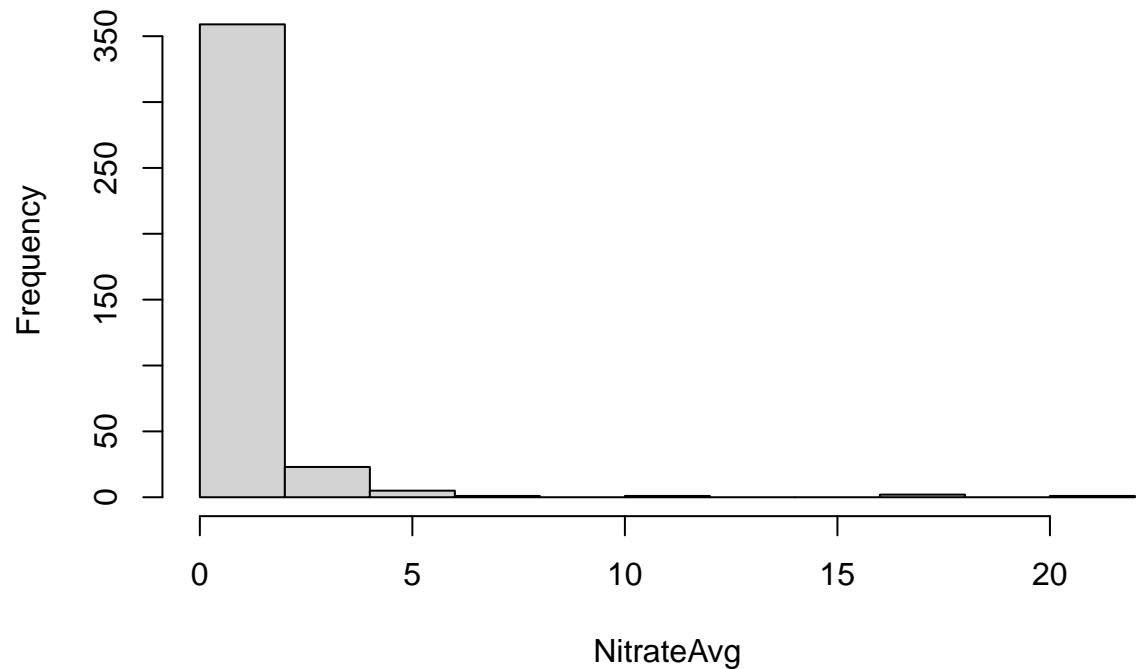




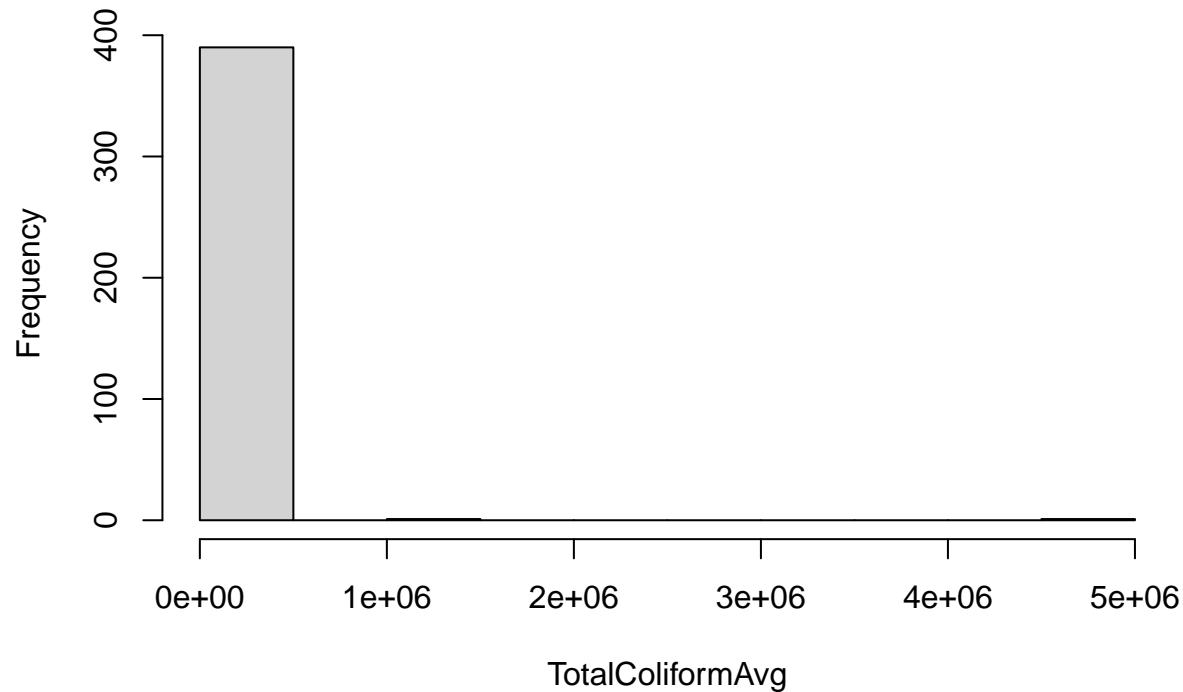
FecalColiformAvg

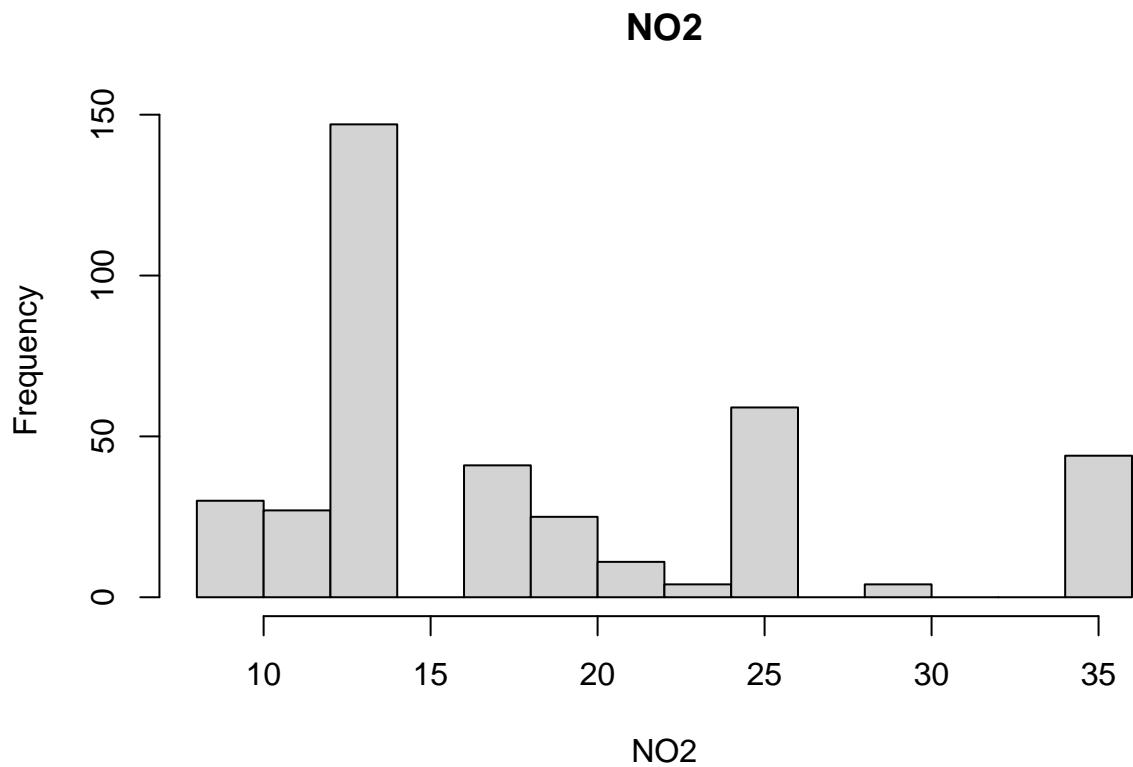


NitrateAvg

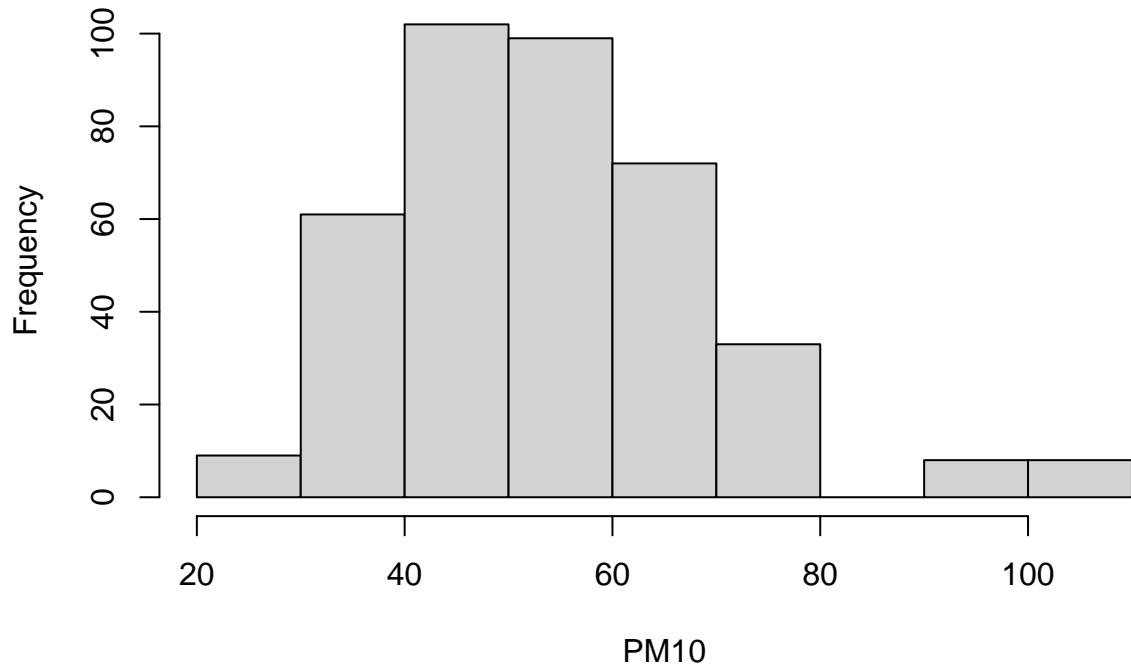


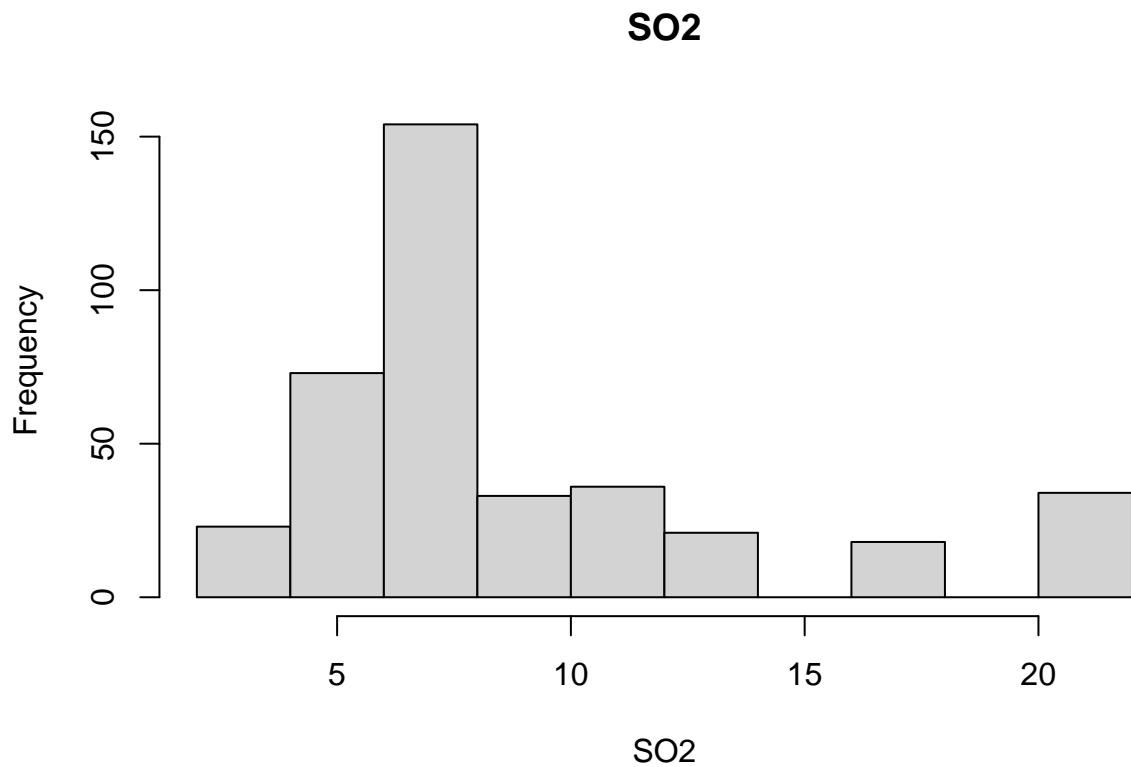
TotalColiformAvg

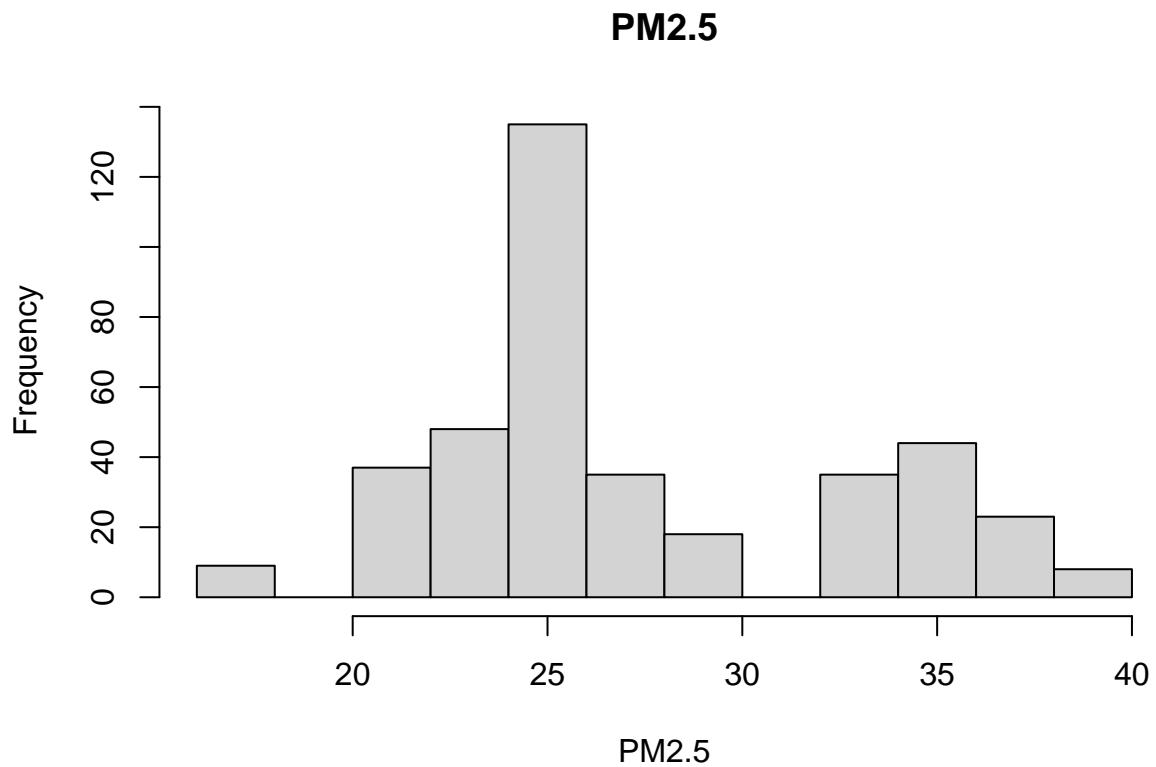




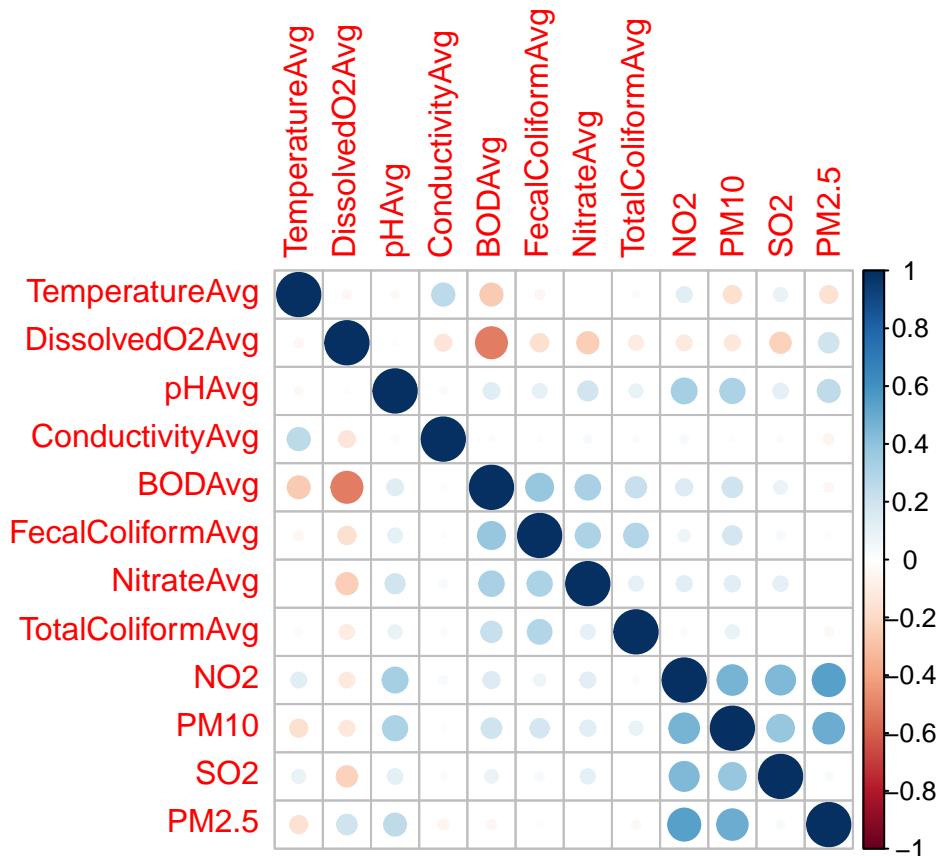
PM10







```
cov_mat = cov(waterQuality_2020_Avg[, numCols])  
  
corr_matrix = cor(waterQuality_2020_Avg[, numCols])  
  
library(corrplot)  
  
corrplot(corr_matrix)
```



```
waterQuality_2020_Avg = aggregate(cbind(TemperatureAvg,
                                         DissolvedO2Avg,
                                         pHAvg,
                                         ConductivityAvg,
                                         BODAvg,
                                         NitrateAvg,
                                         TotalColiformAvg,
                                         FecalColiformAvg,
                                         NO2,
                                         PM10,
                                         SO2,
                                         PM2.5) ~ State,
                                     data = waterQuality_2020_Avg,
                                     mean)
```

```
pca_data = waterQuality_2020_Avg[, c("TemperatureAvg"
, "DissolvedO2Avg"
, "pHAvg"
, "ConductivityAvg"
, "BODAvg"
, "NitrateAvg"
, "FecalColiformAvg"
, "TotalColiformAvg"
, "NO2"
, "PM10"
, "SO2"]
```

```

,"PM2.5")]

# Center and scale the data
pca_data_scaled = scale(pca_data)

# Perform PCA
pca_result = princomp(pca_data_scaled, center = TRUE, scale. = TRUE)

## Warning: In princomp.default(pca_data_scaled, center = TRUE, scale. = TRUE) :
##   extra arguments 'center', 'scale.' will be disregarded

# Print the summary of PCA results
summary(pca_result)

## Importance of components:
##                               Comp.1    Comp.2    Comp.3    Comp.4    Comp.5
## Standard deviation      1.9806715 1.4406720 1.2680676 1.0803567 0.91731267
## Proportion of Variance 0.3432677 0.1816094 0.1406996 0.1021274 0.07362797
## Cumulative Proportion  0.3432677 0.5248771 0.6655767 0.7677041 0.84133208
##                               Comp.6    Comp.7    Comp.8    Comp.9    Comp.10
## Standard deviation     0.80442642 0.59145607 0.5211887 0.45981530 0.44478427
## Proportion of Variance 0.05662141 0.03060927 0.0237683 0.01850013 0.01731039
## Cumulative Proportion  0.89795350 0.92856277 0.9523311 0.97083120 0.98814159
##                               Comp.11   Comp.12
## Standard deviation     0.305279536 0.205740221
## Proportion of Variance 0.008154615 0.003703791
## Cumulative Proportion  0.996296209 1.000000000

#reverse the signs
pca_result$rotation = -1*pca_result$rotation

#display principal components
pca_result$rotation

## numeric(0)

var_explained = pca_result$sdev^2 / sum(pca_result$sdev^2)

print(var_explained)

##          Comp.1    Comp.2    Comp.3    Comp.4    Comp.5    Comp.6
## 0.343267702 0.181609393 0.140699593 0.102127422 0.073627972 0.056621413
##          Comp.7    Comp.8    Comp.9    Comp.10   Comp.11   Comp.12
## 0.030609274 0.023768298 0.018500135 0.017310392 0.008154615 0.003703791

pc.comp = pca_result$scores

#85%
pc.comp1 = -1*pc.comp[,1]
pc.comp2 = -1*pc.comp[,2]

```

```

pc.comp3 = -1*pc.comp[,3]
pc.comp4 = -1*pc.comp[,4]
pc.comp5 = -1*pc.comp[,5]

library(factoextra)
library(tidyr)
library(ggplot2)

X = cbind(pc.comp1, pc.comp2, pc.comp3, pc.comp4, pc.comp5)

X

##          pc.comp1    pc.comp2    pc.comp3    pc.comp4    pc.comp5
## [1,] -1.2851263 -1.603192702 -0.4016635 -0.03474576 -1.19930610
## [2,] -0.9138134 -0.108013419  1.0875392  0.06643863 -0.15829311
## [3,]  1.7175469  2.141605360  0.9108770 -1.53822055  0.29335600
## [4,] -0.3009693 -0.908990601 -1.5576557 -0.46685337 -0.74984955
## [5,] -1.9893890 -1.596157639 -1.3232075 -0.26714779  0.74629048
## [6,]  0.5153036  0.214451174 -1.8688284  1.01657301  0.93632983
## [7,]  6.0970644 -1.832311523 -0.1829389 -0.89976873 -1.92133652
## [8,] -1.0818856  1.718973391  2.3722424 -2.04953517  0.01046893
## [9,]  1.6856088  3.162821763 -0.6585425  1.81948042  0.56017980
## [10,] -1.5881729 -0.120274458  0.5537967  0.16851132 -0.82377412
## [11,] -1.9339509 -1.110578970  0.3112428 -0.23962803  0.17666075
## [12,] -0.7879074 -0.093031918 -0.2582152  0.74464123 -1.05476673
## [13,]  0.1194263  1.581320396 -1.8163063 -0.04424074 -0.05728972
## [14,] -0.6922020  0.610882466  0.6766816 -0.10324749 -0.20513970
## [15,] -1.2687802 -0.859114341  2.4786095  0.60758067  0.19117899
## [16,] -1.1084725 -0.939002621 -2.0654284 -2.26884560  1.53085453
## [17,]  4.0150105 -2.370852652  1.5406457  0.93129527  2.31689262
## [18,]  1.4912822  2.295692516 -0.5293420 -0.50194919 -0.10700039
## [19,] -0.5436541 -0.202156507  0.1758895  0.10007912  0.05094236
## [20,] -1.4531568  0.010012591  0.6983938  0.84071992  0.14785912
## [21,] -0.6937623  0.007917696 -0.1437899  2.11886283 -0.68425748

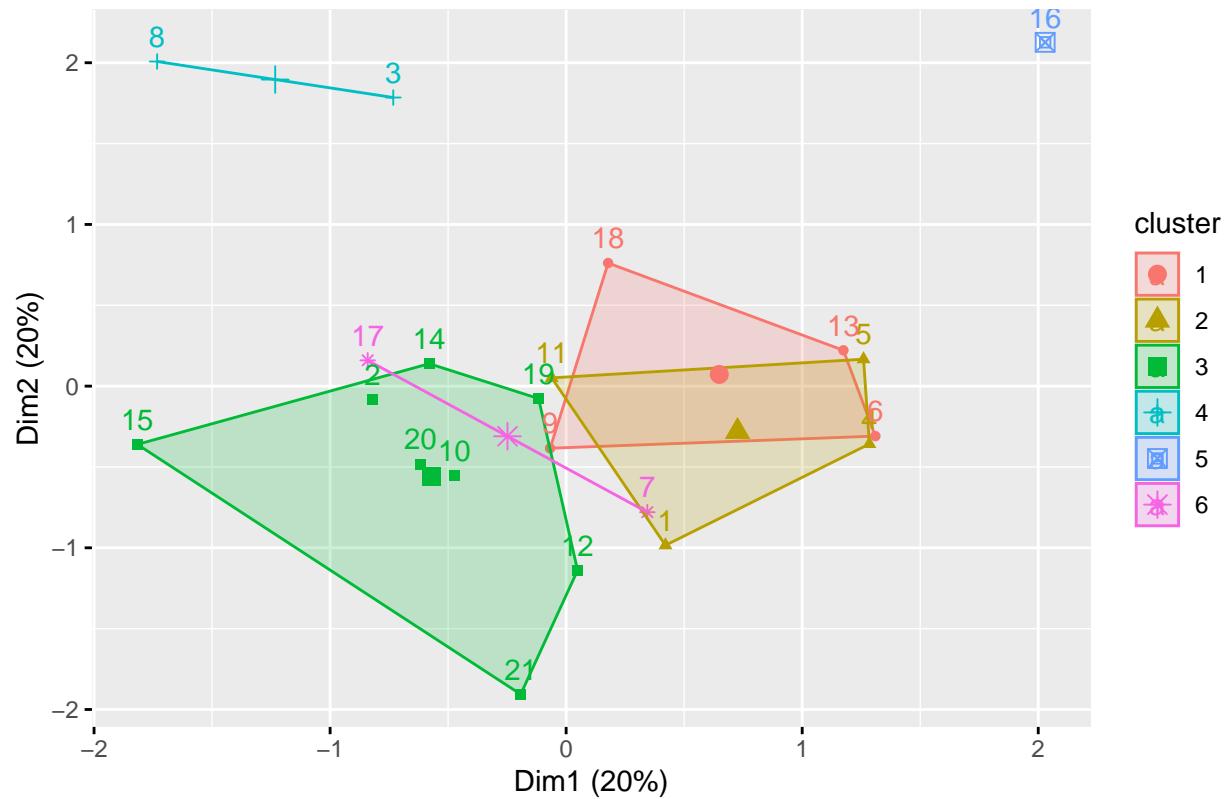
k = 6
kmeans_model = kmeans(X, centers = k, nstart=1)

waterQuality_2020_Avg$cluster=kmeans_model$cluster

fviz_cluster(kmeans_model, data = X)

```

Cluster plot



```
mean_by_group = aggregate(cbind(TemperatureAvg, DissolvedO2Avg, pHAvg, ConductivityAvg, BODAvg, NitrateAvg),
table(waterQuality_2020_Avg$State, waterQuality_2020_Avg$cluster)
```

```
##
##          1 2 3 4 5 6
## ANDHRA PRADESH 0 1 0 0 0 0
## ASSAM        0 0 1 0 0 0
## BIHAR         0 0 0 1 0 0
## DAMAN AND DIU, \r\nDADRA AND\r\nNAGAR HAVELI 0 1 0 0 0 0
## GOA          0 1 0 0 0 0
## GUJARAT      1 0 0 0 0 0
## HARYANA      0 0 0 0 0 1
## HIMACHAL\r\nPRADESH 0 0 0 1 0 0
## JHARKHAND    1 0 0 0 0 0
## KARNATAKA    0 0 1 0 0 0
## KERALA        0 1 0 0 0 0
## MADHYA PRADESH 0 0 1 0 0 0
## MAHARASHTRA   1 0 0 0 0 0
## MANIPUR       0 0 1 0 0 0
## MEGHALAYA    0 0 1 0 0 0
## PUDUCHERRY   0 0 0 0 1 0
## PUNJAB        0 0 0 0 0 1
## RAJASTHAN     1 0 0 0 0 0
## State Name    0 0 1 0 0 0
```

```

##    TAMIL NADU          0 0 1 0 0 0
##    TRIPURA            0 0 1 0 0 0

```

waterQuality_2020_Avg

		State	TemperatureAvg	DissolvedO2Avg			
##	1	ANDHRA PRADESH	24.11111	6.422222			
##	2	ASSAM	19.36364	6.368182			
##	3	BIHAR	17.00000	6.537500			
##	4	DAMAN AND DIU, \r\nDADRA AND\r\nNAGAR HAVELI	28.20000	5.485000			
##	5	GOA	26.96154	6.263462			
##	6	GUJARAT	23.83333	5.130556			
##	7	HARYANA	19.55820	5.030000			
##	8	HIMACHAL\r\nPRADESH	13.98459	8.536957			
##	9	JHARKHAND	15.12500	6.175000			
##	10	KARNATAKA	24.78571	7.303571			
##	11	KERALA	24.50685	6.441781			
##	12	MADHYA PRADESH	20.81818	6.681818			
##	13	MAHARASHTRA	25.05556	6.551389			
##	14	MANIPUR	18.83871	6.675806			
##	15	MEGHALAYA	13.96667	6.810000			
##	16	PUDUCHERRY	26.00000	5.375000			
##	17	PUNJAB	12.94268	3.369444			
##	18	RAJASTHAN	17.25000	5.675000			
##	19	State Name	21.32275	6.340409			
##	20	TAMIL NADU	23.14286	6.653571			
##	21	TRIPURA	22.15385	6.071154			
##	pHAvg	ConductivityAvg	BODAvg	NitrateAvg	TotalColiformAvg		
##	1	7.716667	4570.8889	2.022222	1.5838889	188.1667	
##	2	7.354545	115.2045	1.788636	0.9340909	1212.6818	
##	3	7.650000	357.0000	2.362500	0.3275000	85950.0000	
##	4	7.620000	4554.4614	2.030000	2.8535000	2800.5422	
##	5	6.955769	11764.1538	1.673077	0.7192308	3935.1154	
##	6	7.594444	6617.8889	11.122222	0.4672222	1794.9957	
##	7	8.020000	1294.5500	22.500000	6.3640000	249930.0000	
##	8	7.667391	429.1957	1.178261	0.7326589	3478.2609	
##	9	7.287500	1589.6195	6.056250	1.1773077	18630.8166	
##	10	7.321429	160.4286	2.325000	0.5121429	660.3214	
##	11	7.003425	4043.2945	1.606849	0.6297945	2820.1164	
##	12	7.804545	2688.3636	1.377273	0.8377273	108.0000	
##	13	7.538889	5099.4028	4.312500	1.2609722	2433.0417	
##	14	7.333871	283.9839	3.403226	1.1030769	229.0484	
##	15	6.781667	171.7833	5.876667	0.9856667	10756.2667	
##	16	7.600000	13066.5000	1.000000	1.3182692	24374.4222	
##	17	7.586111	880.7500	24.638889	1.4944444	307158.3889	
##	18	8.350000	818.7500	1.825000	1.5037500	121.3750	
##	19	7.293606	2956.6140	4.527621	1.1030769	24374.4222	
##	20	6.732143	367.2500	3.600000	0.4275000	177.3571	
##	21	7.113462	146.8077	1.211538	1.3053846	163.6731	
##	FecalColiformAvg		NO2	PM10	S02	PM2.5 cluster	
##	1	22.72222	17.562500	28.25000	7.125000	17.56250	2
##	2	513.63636	13.429487	58.08731	7.089744	25.67907	3
##	3	80997.50000	22.136364	94.04545	8.818922	39.12783	4
##	4	787.21108	18.077739	55.24444	8.949573	27.48306	2

```

## 5      2320.28846 13.411765  59.06424  7.021524 21.26011      2
## 6      545.28393 24.562963  67.99259 16.120823 29.08395      1
## 7    109189.00000 24.576923  78.03846 12.346154 26.32933      6
## 8     1241.78261 11.119048  71.55952  2.226190 36.89954      4
## 9     4225.33311 34.255556 104.85260 21.177778 35.46457      1
## 10    322.35714 17.102624  47.58049  7.824026 24.67611      3
## 11    1888.94521 12.891386  43.55600  4.900582 24.99776      2
## 12     19.77273 20.947368  40.36842 13.243725 20.94737      3
## 13    1322.68056 34.694938  67.94619 11.944126 35.15770      1
## 14     56.51613 24.231229  38.50000  7.254072 32.64061      3
## 15    3377.03333  9.928571  39.37500  6.562912 23.85490      3
## 16     5564.61082 11.50000  41.83333  6.500000 39.28123      5
## 17   16130.61111 17.832208  67.48132  7.546342 23.52170      6
## 18     67.25000 28.032680  99.91756  9.937908 32.46874      1
## 19     5564.61082 18.077739  55.24444  8.949573 27.48306      3
## 20     49.50000 18.077739  55.24444  8.949573 27.48306      3
## 21     43.28846 13.000000  55.75000 21.000000 24.25000      3

```

```

Result_2020<- mean_by_group[order(mean_by_group$TotalColiformAvg, mean_by_group$BODAvg, mean_by_group$Di

Result_2020$Rank=c("A", "B", "C", "D", "E", "F")

class(Result_2020$cluster)

## [1] "integer"

class(waterQuality_2020_Avg$cluster)

## [1] "integer"

WaterAir2020 = merge(waterQuality_2020_Avg, Result_2020[,c("cluster","Rank")], by = "cluster", all.x = T

WaterQualityAlone = waterQuality_2020_Avg[, c("TemperatureAvg"
,"DissolvedO2Avg"
,"pHAvg"
,"ConductivityAvg"
,"BODAvg"
,"FecalColiformAvg"
,"NitrateAvg"
,"TotalColiformAvg"
)]]

# Center and scale the data
WaterQualityAlone_scaled = scale(WaterQualityAlone)

# Perform PCA
pca_result = princomp(WaterQualityAlone_scaled, center = TRUE, scale. = TRUE)

## Warning: In princomp.default(WaterQualityAlone_scaled, center = TRUE, scale. = TRUE) :
## extra arguments 'center', 'scale.' will be disregarded

```

```

# Print the summary of PCA results
summary(pca_result)

## Importance of components:
##                               Comp.1     Comp.2     Comp.3     Comp.4     Comp.5
## Standard deviation      1.8771194 1.2726958 0.9838068 0.82960438 0.62402420
## Proportion of Variance 0.4624695 0.2125928 0.1270337 0.09033195 0.05110956
## Cumulative Proportion  0.4624695 0.6750623 0.8020960 0.89242796 0.94353752
##                               Comp.6     Comp.7     Comp.8
## Standard deviation      0.47911831 0.37045978 0.251784608
## Proportion of Variance 0.03012901 0.01801281 0.008320658
## Cumulative Proportion  0.97366653 0.99167934 1.0000000000

#reverse the signs
pca_result$rotation = -1*pca_result$rotation

#display principal components
pca_result$rotation

## numeric(0)

var_explained = pca_result$sdev^2 / sum(pca_result$sdev^2)

print(var_explained)

##           Comp.1     Comp.2     Comp.3     Comp.4     Comp.5     Comp.6
## 0.462469526 0.212592787 0.127033698 0.090331950 0.051109564 0.030129009
##           Comp.7     Comp.8
## 0.018012809 0.008320658

pc.comp = pca_result$scores

#86%
pc.comp1 = -1*pc.comp[,1]
pc.comp2 = -1*pc.comp[,2]
pc.comp3 = -1*pc.comp[,3]
pc.comp4 = -1*pc.comp[,4]

library(factoextra)
library(tidyverse)
library(ggplot2)

X = cbind(pc.comp1, pc.comp2, pc.comp3, pc.comp4)

X

##          pc.comp1    pc.comp2    pc.comp3    pc.comp4
## [1,] -0.59382232 -0.76925689 -0.772846752  0.387567670
## [2,] -0.66143537  0.84505098  0.020412417  0.127970744
## [3,]  1.01692079  1.31005220 -0.972798278 -0.495115617

```

```

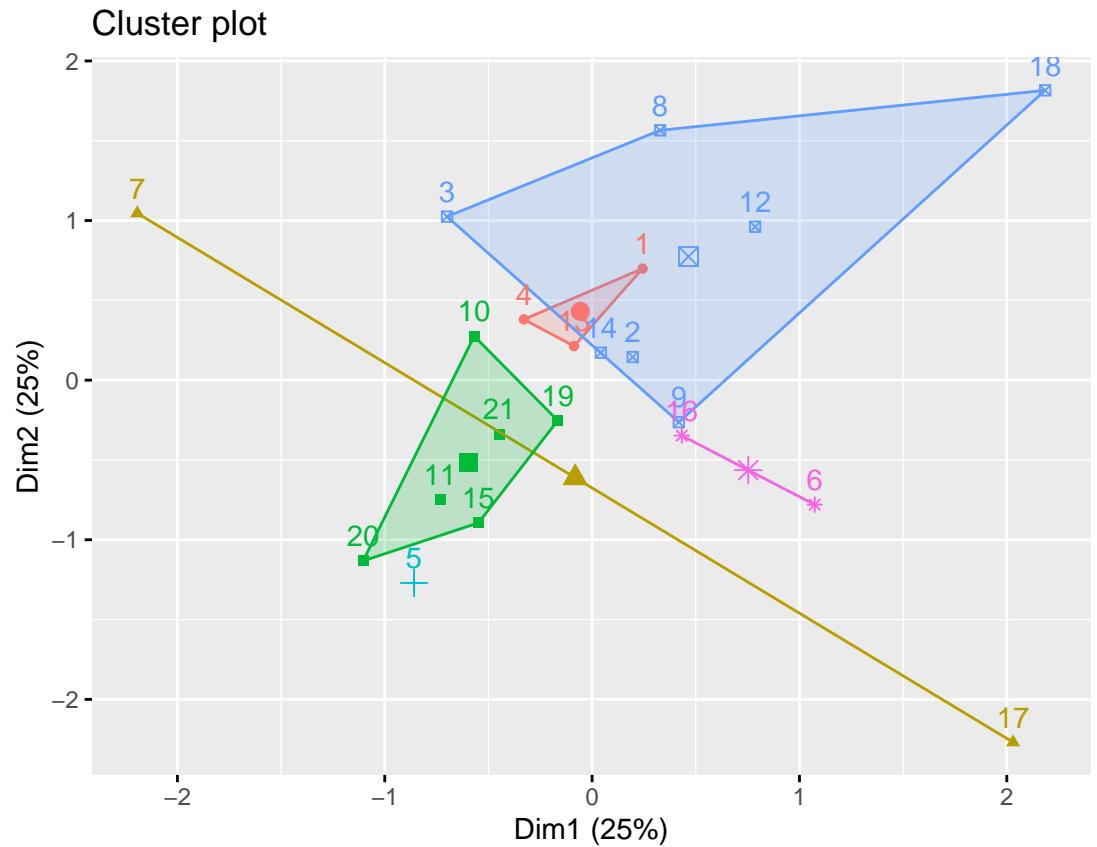
## [4,] -0.09908167 -1.80653118 -0.721192501 -0.072426382
## [5,] -1.59354983 -2.19632599  0.779877532 -0.786923189
## [6,]  0.05972456 -1.36028433  0.792242785  0.855895315
## [7,]  6.23694452 -0.65622353 -1.585274542 -1.594438885
## [8,] -1.07274269  2.22150454 -1.177658587  0.395178519
## [9,]  0.04092148  1.07046771  0.509690234  0.222593203
## [10,] -1.35304852  0.49441055 -0.315154957 -0.461551771
## [11,] -1.33414751 -0.40513278  0.549406960 -0.719231672
## [12,] -0.69795231  0.19661060 -0.784029822  0.817551813
## [13,] -0.71470244 -0.85790635 -0.361812573  0.029591625
## [14,] -0.61018309  0.96942982 -0.018792223 -0.005035955
## [15,] -0.53850118  1.84996560  1.073130654 -0.758119227
## [16,] -0.56918713 -2.84219278  0.006200914  0.508146848
## [17,]  4.59421978  0.47136491  2.758546805  1.249273184
## [18,]  0.45262717  0.44095246 -1.333389764  2.124703385
## [19,] -0.41223652  0.05452284  0.230855995 -0.188709303
## [20,] -1.37260978  0.60786634  1.015840765 -1.174409782
## [21,] -0.77815794  0.36165528  0.306744937 -0.462510523

k =6
kmeans_model = kmeans(X, centers = k,nstart=1)

waterQuality_2020_Avg$cluster=kmeans_model$cluster

fviz_cluster(kmeans_model, data = X)

```



```
mean_by_group = aggregate(cbind(TemperatureAvg, DissolvedO2Avg, pHAvg, ConductivityAvg, BODAvg, NitrateAvg),
                          by = list(waterQuality_2020_Avg$State), FUN = mean)
table(waterQuality_2020_Avg$State, waterQuality_2020_Avg$cluster)
```

```
##                                     1 2 3 4 5 6
## ANDHRA PRADESH                 1 0 0 0 0 0
## ASSAM                         0 0 0 0 1 0
## BIHAR                          0 0 0 0 1 0
## DAMAN AND DIU, \r\nDADRA AND\r\nNAGAR HAVELI 1 0 0 0 0 0
## GOA                            0 0 0 1 0 0
## GUJARAT                        0 0 0 0 0 1
## HARYANA                         0 1 0 0 0 0
## HIMACHAL\r\nPRADESH              0 0 0 0 1 0
## JHARKHAND                       0 0 0 0 1 0
## KARNATAKA                       0 0 1 0 0 0
## KERALA                           0 0 1 0 0 0
## MADHYA PRADESH                  0 0 0 0 1 0
## MAHARASHTRA                      1 0 0 0 0 0
## MANIPUR                          0 0 0 0 1 0
## MEGHALAYA                        0 0 1 0 0 0
## PUDUCHERRY                      0 0 0 0 0 1
## PUNJAB                           0 1 0 0 0 0
## RAJASTHAN                         0 0 0 0 1 0
## State Name                       0 0 1 0 0 0
```

```

##    TAMIL NADU          0 0 1 0 0 0
##    TRIPURA            0 0 1 0 0 0

```

WaterQualityAlone

```

##    TemperatureAvg DissolvedO2Avg    pHAvg ConductivityAvg    BODAvg
## 1      24.11111     6.422222 7.716667      4570.8889  2.022222
## 2      19.36364     6.368182 7.354545      115.2045  1.788636
## 3      17.00000     6.537500 7.650000      357.0000  2.362500
## 4      28.20000     5.485000 7.620000      4554.4614  2.030000
## 5      26.96154     6.263462 6.955769     11764.1538  1.673077
## 6      23.83333     5.130556 7.594444      6617.8889 11.122222
## 7      19.55820     5.030000 8.020000      1294.5500 22.500000
## 8      13.98459     8.536957 7.667391      429.1957  1.178261
## 9      15.12500     6.175000 7.287500      1589.6195  6.056250
## 10     24.78571     7.303571 7.321429      160.4286  2.325000
## 11     24.50685     6.441781 7.003425      4043.2945  1.606849
## 12     20.81818     6.681818 7.804545      2688.3636  1.377273
## 13     25.05556     6.551389 7.538889      5099.4028  4.312500
## 14     18.83871     6.675806 7.333871      283.9839  3.403226
## 15     13.96667     6.810000 6.781667      171.7833  5.876667
## 16     26.00000     5.375000 7.600000     13066.5000  1.000000
## 17     12.94268     3.369444 7.586111      880.7500 24.638889
## 18     17.25000     5.675000 8.350000      818.7500  1.825000
## 19     21.32275     6.340409 7.293606      2956.6140  4.527621
## 20     23.14286     6.653571 6.732143      367.2500  3.600000
## 21     22.15385     6.071154 7.113462      146.8077  1.211538
##    FecalColiformAvg NitrateAvg TotalColiformAvg
## 1      22.72222   1.5838889      188.1667
## 2      513.63636   0.9340909      1212.6818
## 3      80997.50000  0.3275000     85950.0000
## 4      787.21108   2.8535000      2800.5422
## 5      2320.28846   0.7192308      3935.1154
## 6      545.28393   0.4672222      1794.9957
## 7      109189.00000  6.3640000     249930.0000
## 8      1241.78261   0.7326589      3478.2609
## 9      4225.33311   1.1773077      18630.8166
## 10     322.35714   0.5121429      660.3214
## 11     1888.94521   0.6297945      2820.1164
## 12     19.77273   0.8377273      108.0000
## 13     1322.68056   1.2609722      2433.0417
## 14     56.51613   1.1030769      229.0484
## 15     3377.03333   0.9856667      10756.2667
## 16     5564.61082   1.3182692      24374.4222
## 17     16130.61111   1.4944444     307158.3889
## 18     67.25000   1.5037500      121.3750
## 19     5564.61082   1.1030769      24374.4222
## 20     49.50000   0.4275000      177.3571
## 21     43.28846   1.3053846      163.6731

```

```

Result_2020_water= mean_by_group[order(mean_by_group$TotalColiformAvg, mean_by_group$BODAvg, mean_by_gr

```

```

Result_2020_water$Rank=c("A", "B", "C", "D", "E", "F")
class(waterQuality_2020_Avg$cluster)

## [1] "integer"

class(waterQuality_2020_Avg$cluster)

## [1] "integer"

Water2020 = merge(waterQuality_2020_Avg, Result_2020_water[,c("cluster","Rank")], by = "cluster", all.x=TRUE)
waterQuality_2021=finalDataset[finalDataset$year == 2021,]
waterQuality_2021$TemperatureMin      =as.numeric(waterQuality_2021$TemperatureMin)

## Warning: NAs introduced by coercion

waterQuality_2021$TemperatureMax      =as.numeric(waterQuality_2021$TemperatureMin)
waterQuality_2021$DissolvedO2Min     =as.numeric(waterQuality_2021$DissolvedO2Min)

## Warning: NAs introduced by coercion

waterQuality_2021$DissolvedO2Max     =as.numeric(waterQuality_2021$DissolvedO2Max)

## Warning: NAs introduced by coercion

waterQuality_2021$pHMin              =as.numeric(waterQuality_2021$pHMin)

## Warning: NAs introduced by coercion

waterQuality_2021$pHMax              =as.numeric(waterQuality_2021$pHMax)

## Warning: NAs introduced by coercion

waterQuality_2021$ConductivityMin   =as.numeric(waterQuality_2021$ConductivityMin)

## Warning: NAs introduced by coercion

waterQuality_2021$ConductivityMax   =as.numeric(waterQuality_2021$ConductivityMax)

## Warning: NAs introduced by coercion

waterQuality_2021$BODMin            =as.numeric(waterQuality_2021$BODMin)

## Warning: NAs introduced by coercion

```

```

waterQuality_2021$BODMax           =as.numeric(waterQuality_2021$BODMax)

## Warning: NAs introduced by coercion

waterQuality_2021$NitrateMin       =as.numeric(waterQuality_2021$NitrateMin)

## Warning: NAs introduced by coercion

waterQuality_2021$NitrateMax       =as.numeric(waterQuality_2021$NitrateMax)

## Warning: NAs introduced by coercion

waterQuality_2021$FecalColiformMin =as.numeric(waterQuality_2021$FecalColiformMin)

## Warning: NAs introduced by coercion

waterQuality_2021$FecalColiformMax =as.numeric(waterQuality_2021$FecalColiformMax)

## Warning: NAs introduced by coercion

waterQuality_2021$TotalColiformMin =as.numeric(waterQuality_2021$TotalColiformMin)

## Warning: NAs introduced by coercion

waterQuality_2021$TotalColiformMax =as.numeric(waterQuality_2021$TotalColiformMax)

## Warning: NAs introduced by coercion

waterQuality_2021$NO2              =as.numeric(waterQuality_2021$NO2)
waterQuality_2021$PM10             =as.numeric(waterQuality_2021$PM10)

cols=c("TemperatureMin"
,"TemperatureMax"
,"DissolvedO2Min"
,"DissolvedO2Max"
,"pHMin"
,"pHMax"
,"ConductivityMin"
,"ConductivityMax"
,"BODMin"
,"BODMax"
,"NitrateMin"
,"NitrateMax"
,"FecalColiformMin"
,"FecalColiformMax"
,"TotalColiformMin"
,"TotalColiformMax"
,"NO2"
,
```

```

,"PM10"
,"SO2"
,"PM2.5")

waterQuality_2021_Avg=waterQuality_2021

# Calculate the average values for each variable
waterQuality_2021_Avg$TemperatureAvg = (waterQuality_2021_Avg$TemperatureMin + waterQuality_2021_Avg$TemperatureMax) / 2
waterQuality_2021_Avg$DissolvedO2Avg = (waterQuality_2021_Avg$DissolvedO2Min + waterQuality_2021_Avg$DissolvedO2Max) / 2
waterQuality_2021_Avg$pHAvg = (waterQuality_2021_Avg$pHMin + waterQuality_2021_Avg$pHMax) / 2
waterQuality_2021_Avg$ConductivityAvg = (waterQuality_2021_Avg$ConductivityMin + waterQuality_2021_Avg$ConductivityMax) / 2
waterQuality_2021_Avg$FecalColiformAvg = (waterQuality_2021_Avg$FecalColiformMin + waterQuality_2021_Avg$FecalColiformMax) / 2
waterQuality_2021_Avg$BODAvg = (waterQuality_2021_Avg$BODMin + waterQuality_2021_Avg$BODMax) / 2
waterQuality_2021_Avg$NitrateAvg = (waterQuality_2021_Avg$NitrateMin + waterQuality_2021_Avg$NitrateMax) / 2

waterQuality_2021_Avg$TotalColiformAvg = (waterQuality_2021_Avg$TotalColiformMin + waterQuality_2021_Avg$TotalColiformMax) / 2

waterQuality_2021_Avg = subset(waterQuality_2021_Avg, select=-c(TemperatureMin
, TemperatureMax
, DissolvedO2Min
, DissolvedO2Max
, pHMin
, pHMax
, ConductivityMin
, ConductivityMax
, BODMin
, BODMax
, NitrateMin
, NitrateMax
, FecalColiformMin
, FecalColiformMax
, TotalColiformMin
, TotalColiformMax))

numCols=c("TemperatureAvg"
,"DissolvedO2Avg"
,"pHAvg"
,"ConductivityAvg"
,"BODAvg"
,"NitrateAvg"
,"TotalColiformAvg"
,"NO2"
,"PM10"
,"SO2"
,"PM2.5")

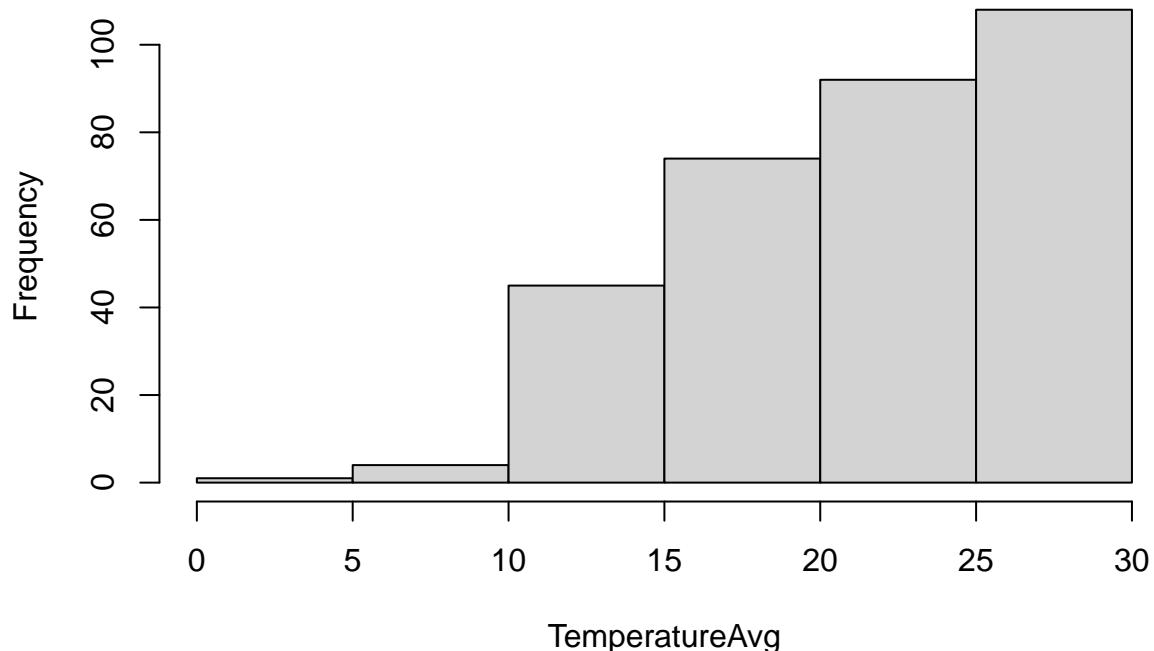
for (col in colnames(waterQuality_2021_Avg)) {
  # Check if column has any NAs
  if (sum(is.na(waterQuality_2021_Avg[, col])) > 0) {
    # Replace NAs with mean value of the column
    waterQuality_2021_Avg[is.na(waterQuality_2021_Avg[, col]), col] = mean(waterQuality_2021_Avg[, col])
  }
}

```

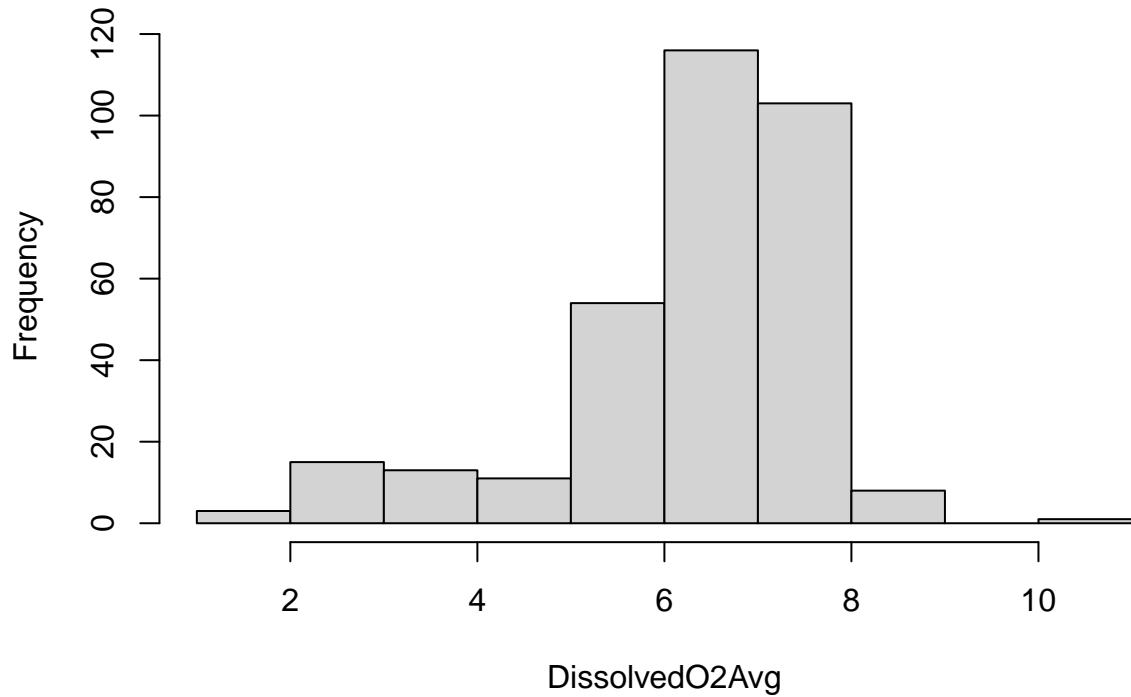
```
}

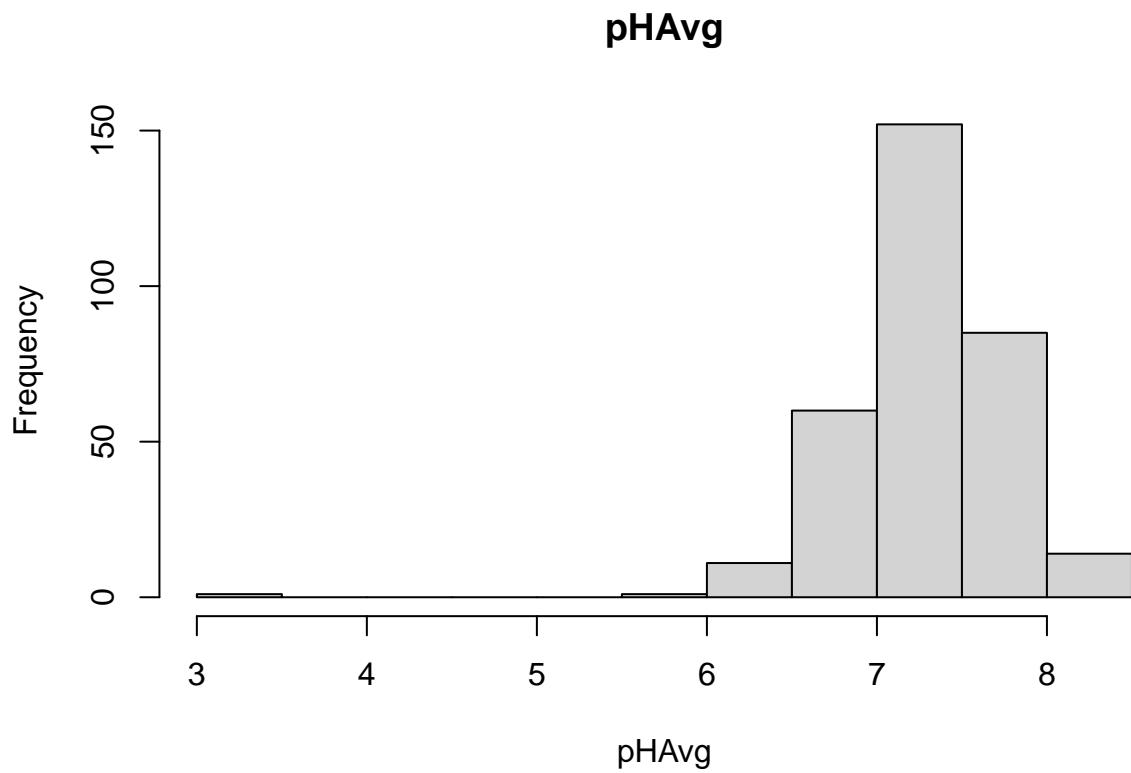
for (col in numCols) {
  # Create the histogram plot
  hist(waterQuality_2021_Avg[[col]], main = col, xlab = col)
}
```

TemperatureAvg

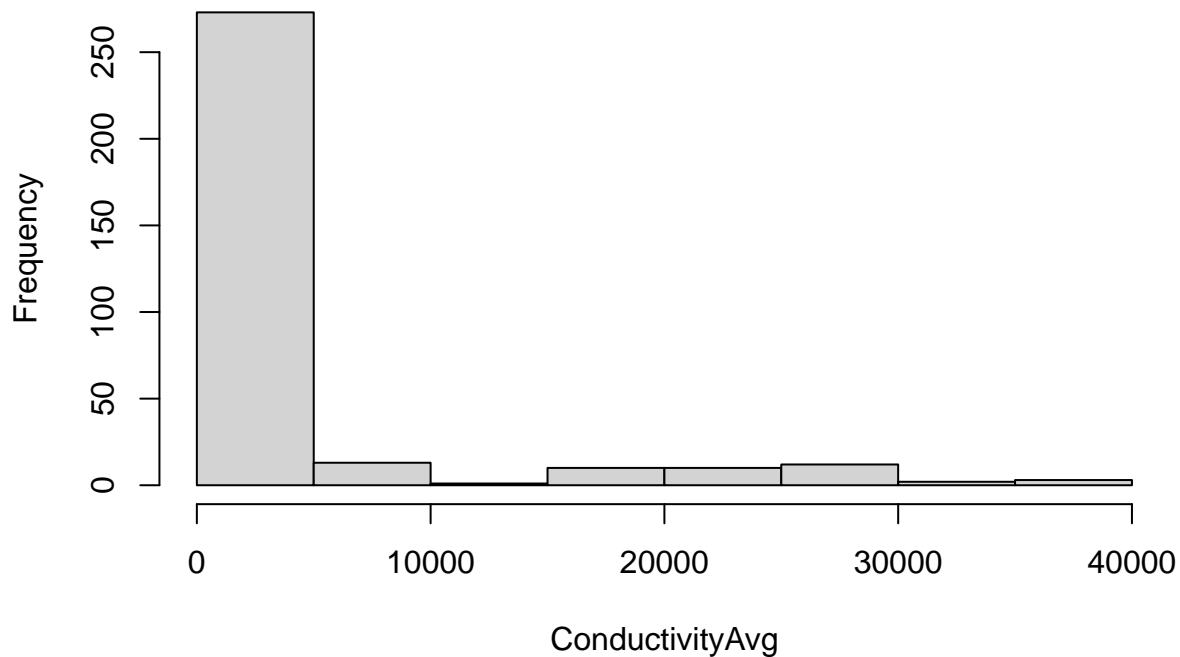


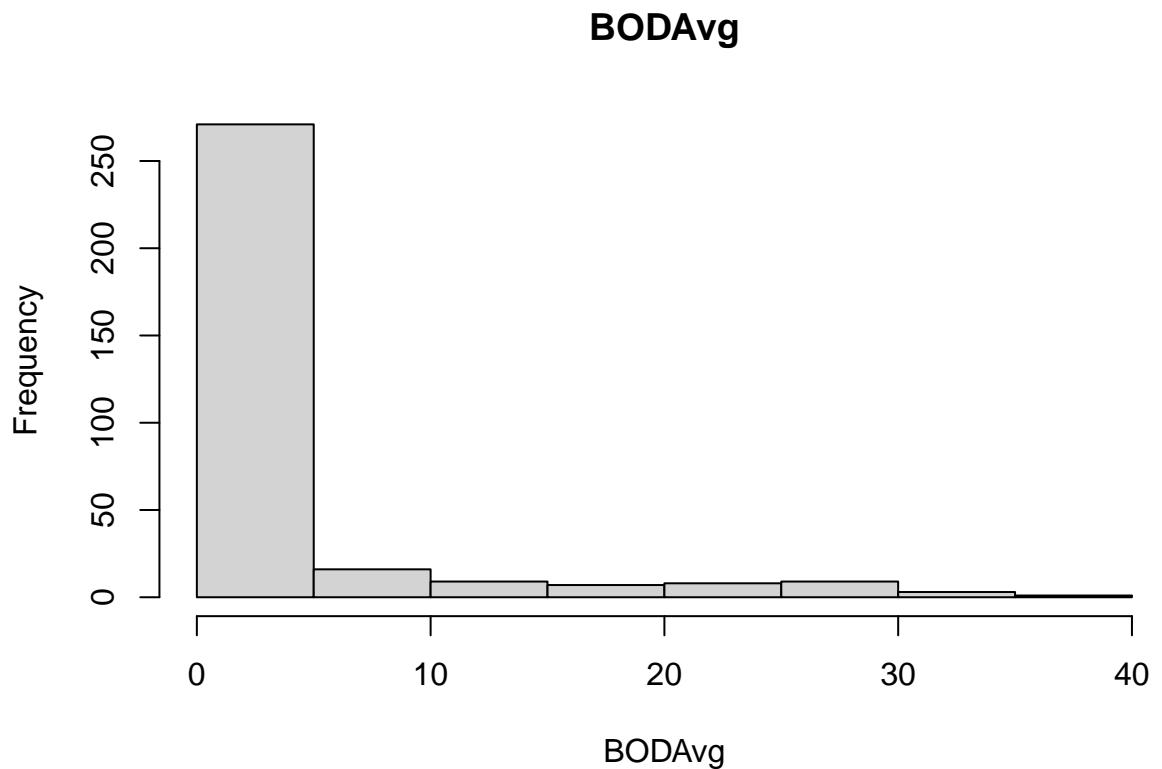
DissolvedO2Avg

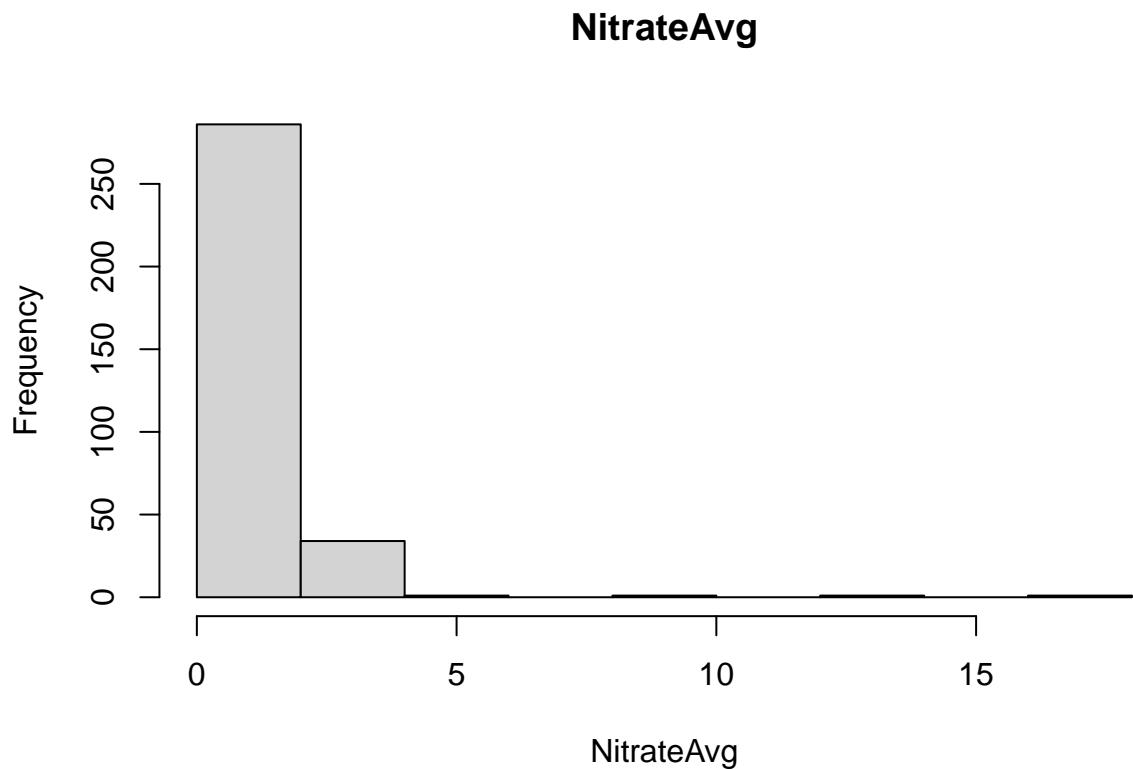




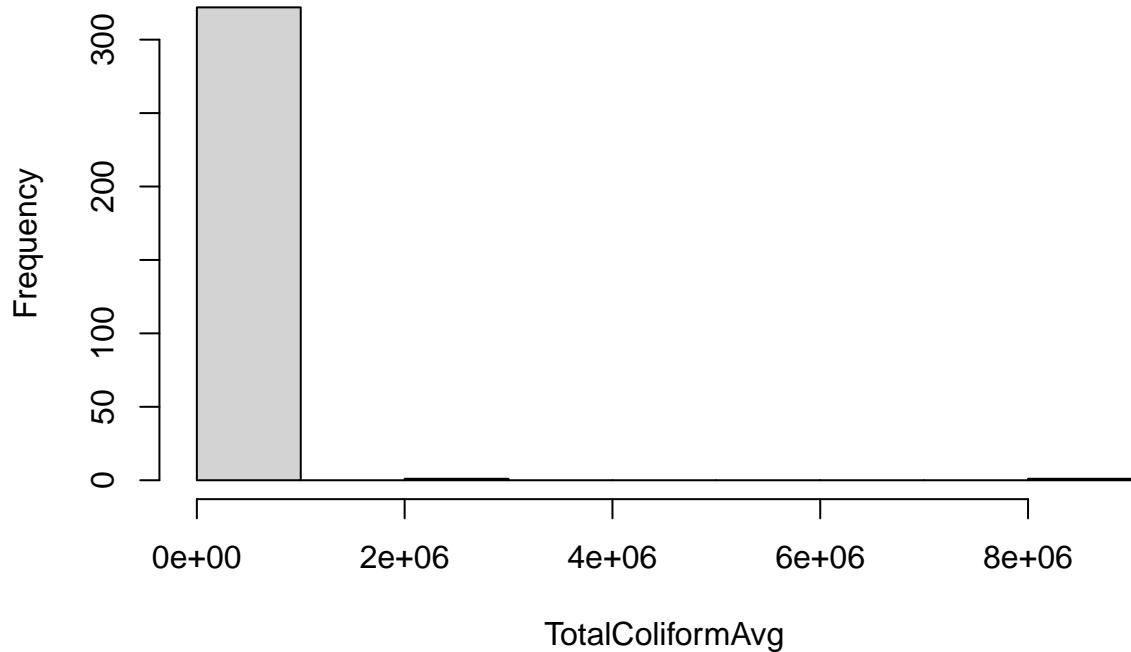
ConductivityAvg



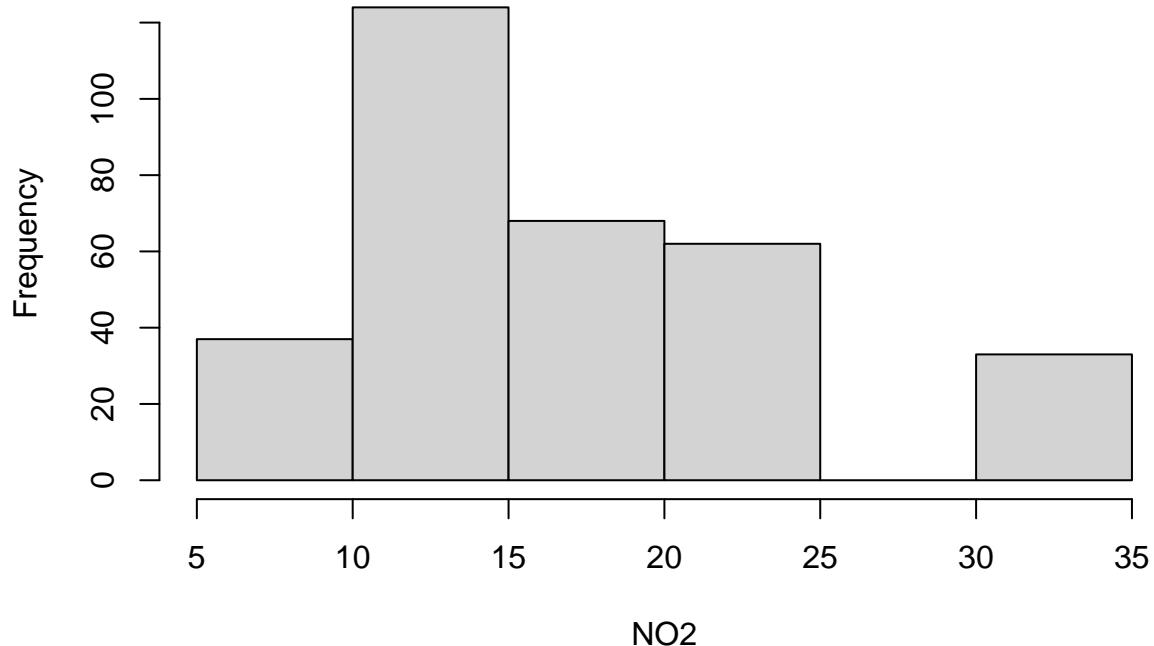




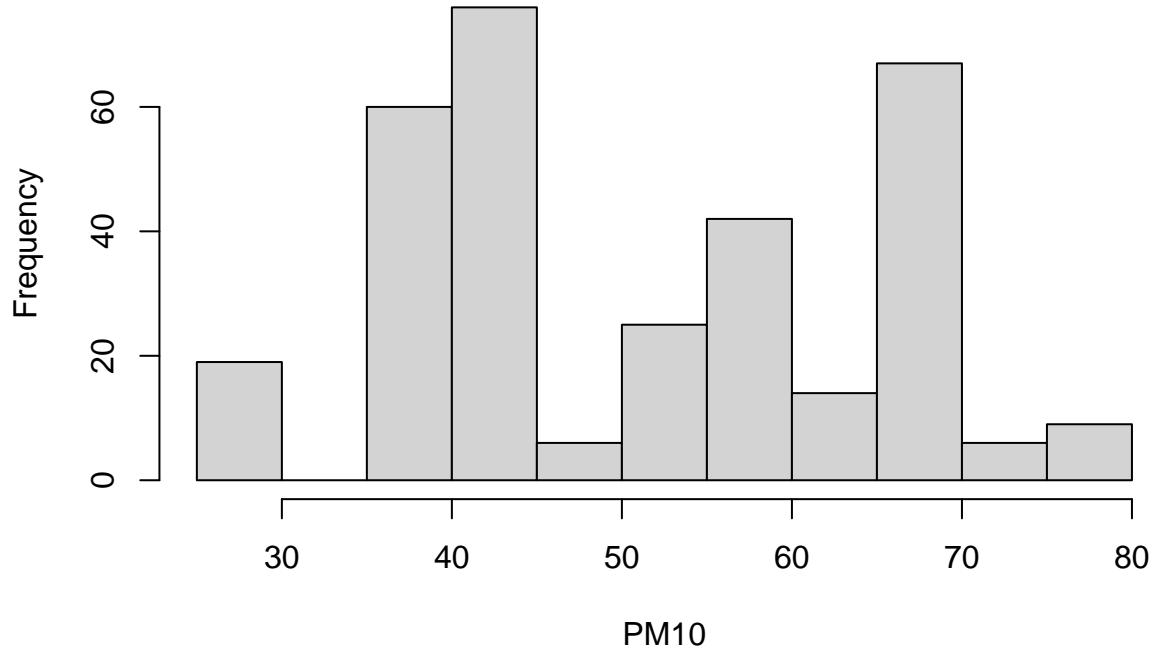
TotalColiformAvg



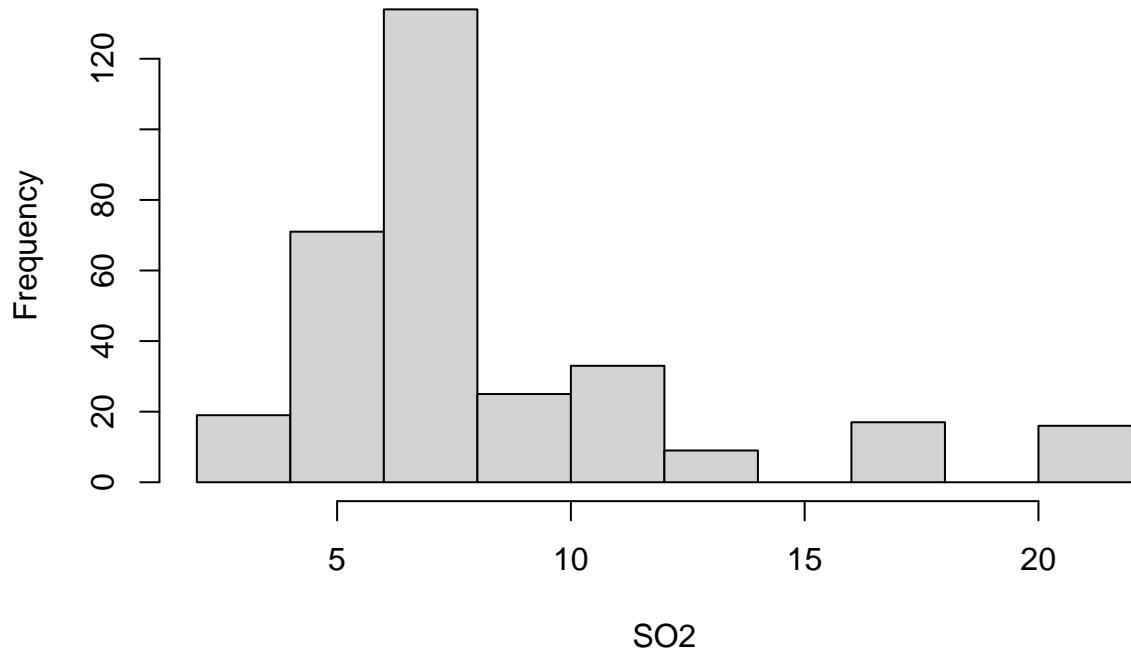
NO2

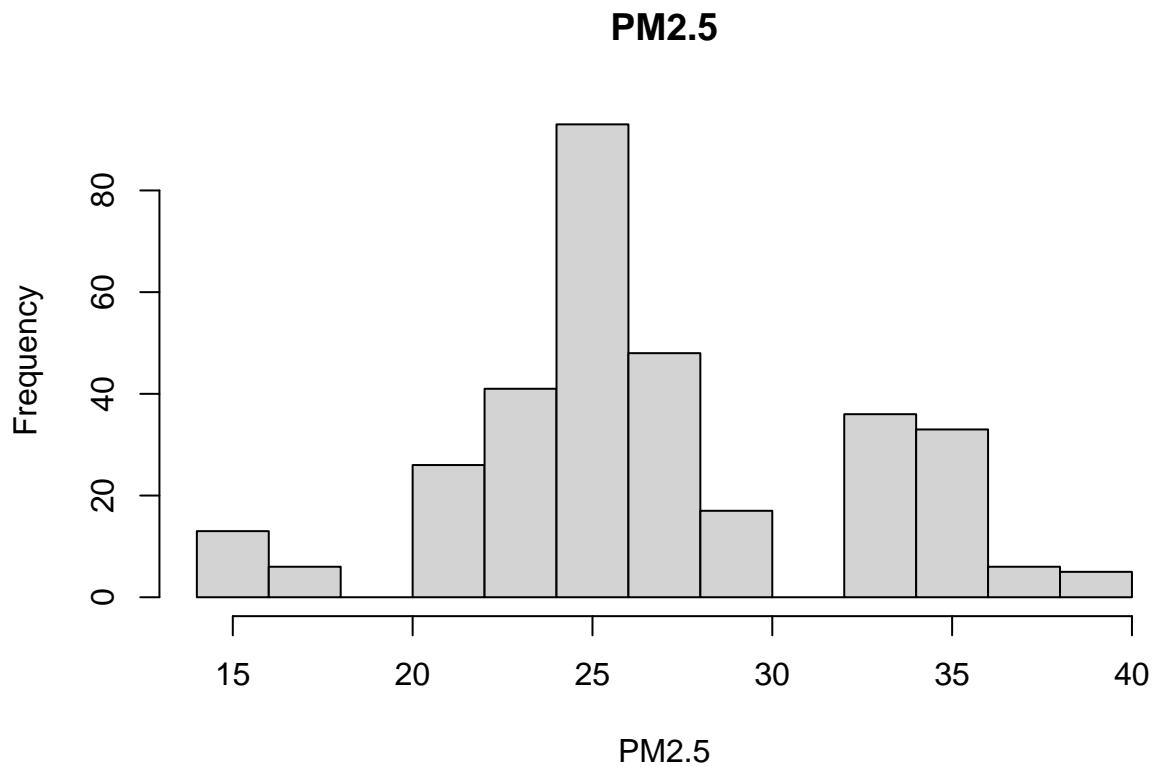


PM10

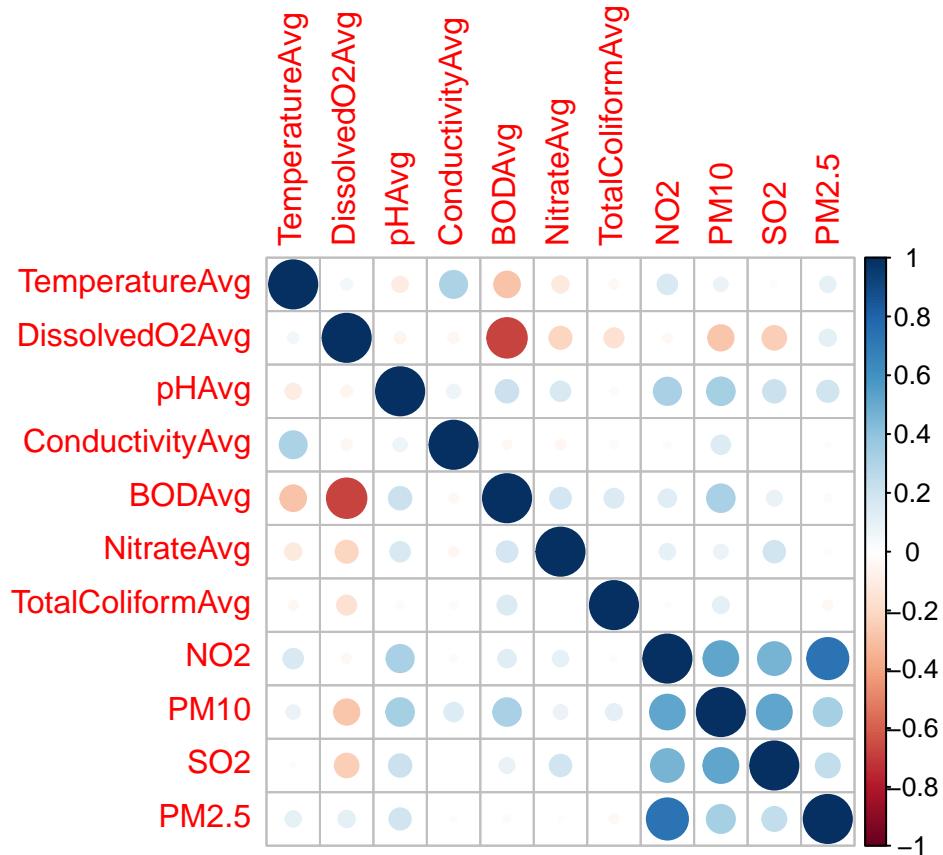


SO₂





```
cov_mat = cov(waterQuality_2021_Avg[, numCols])  
  
corr_matrix = cor(waterQuality_2021_Avg[, numCols])  
  
library(corrplot)  
  
corrplot(corr_matrix)
```



```
waterQuality_2021_Avg = aggregate(cbind(TemperatureAvg,
                                         DissolvedO2Avg,
                                         pHAvg,
                                         ConductivityAvg,
                                         BODAvg,
                                         NitrateAvg,
                                         TotalColiformAvg,
                                         FecalColiformAvg,
                                         NO2,
                                         PM10,
                                         SO2,
                                         PM2.5) ~ State,
                                     data = waterQuality_2021_Avg,
                                     mean)
```

```
pca_data = waterQuality_2021_Avg[, c("TemperatureAvg"
                                         , "DissolvedO2Avg"
                                         , "pHAvg"
                                         , "ConductivityAvg"
                                         , "BODAvg"
                                         , "NitrateAvg"
                                         , "FecalColiformAvg"
                                         , "TotalColiformAvg"
                                         , "NO2"
                                         , "PM10"
                                         , "SO2")]
```

```

,"PM2.5")]

# Center and scale the data
pca_data_scaled = scale(pca_data)

# Perform PCA
pca_result = princomp(pca_data_scaled, center = TRUE, scale. = TRUE)

## Warning: In princomp.default(pca_data_scaled, center = TRUE, scale. = TRUE) :
##   extra arguments 'center', 'scale.' will be disregarded

# Print the summary of PCA results
summary(pca_result)

## Importance of components:
##                               Comp.1    Comp.2    Comp.3    Comp.4    Comp.5
## Standard deviation      1.9376299 1.4515835 1.2424905 1.0652976 0.92710471
## Proportion of Variance 0.3312714 0.1859201 0.1362161 0.1001346 0.07584028
## Cumulative Proportion  0.3312714 0.5171916 0.6534077 0.7535423 0.82938256
##                               Comp.6    Comp.7    Comp.8    Comp.9    Comp.10
## Standard deviation      0.7818993 0.68406166 0.61860516 0.55749759 0.333196286
## Proportion of Variance 0.0539441 0.04128885 0.03376521 0.02742384 0.009795862
## Cumulative Proportion  0.8833267 0.92461551 0.95838072 0.98580456 0.995600426
##                               Comp.11   Comp.12
## Standard deviation      0.186938080 0.122131030
## Proportion of Variance 0.003083457 0.001316117
## Cumulative Proportion  0.998683883 1.000000000

#reverse the signs
pca_result$rotation = -1*pca_result$rotation

#display principal components
pca_result$rotation

## numeric(0)

var_explained = pca_result$sdev^2 / sum(pca_result$sdev^2)

print(var_explained)

##      Comp.1    Comp.2    Comp.3    Comp.4    Comp.5    Comp.6
## 0.331271449 0.185920105 0.136216118 0.100134608 0.075840278 0.053944101
##      Comp.7    Comp.8    Comp.9    Comp.10   Comp.11   Comp.12
## 0.041288855 0.033765207 0.027423843 0.009795862 0.003083457 0.001316117

pc.comp = pca_result$scores

#85%
pc.comp1 = -1*pc.comp[,1]
pc.comp2 = -1*pc.comp[,2]

```

```

pc.comp3 = -1*pc.comp[,3]
pc.comp4 = -1*pc.comp[,4]
pc.comp5 = -1*pc.comp[,5]

library(factoextra)
library(tidyr)
library(ggplot2)

X = cbind(pc.comp1, pc.comp2, pc.comp3, pc.comp4, pc.comp5)

##          pc.comp1    pc.comp2    pc.comp3    pc.comp4    pc.comp5
## [1,] -1.2256159  1.21290290 -1.09905181 -0.26748734 -1.95934192
## [2,]  0.1361639 -0.44459065 -1.35355321 -0.38420450 -1.82227074
## [3,] -1.3054933 -1.34471786  1.24277836 -2.05127525  0.38093732
## [4,]  1.1320970 -2.26109729 -0.55426752 -0.68091507  0.32014166
## [5,]  5.5096568  0.08379829  0.37463154  0.38626322 -0.68143113
## [6,] -0.8799462 -0.22837961  1.10496637  2.74552006  0.44619980
## [7,] -1.2101904  0.51312963 -1.28112658 -0.01164555 -0.20454280
## [8,] -1.7484002  0.02625637  0.95987475 -0.86039828  0.67342956
## [9,]  0.7555221 -2.33286275 -1.78801173  0.96761355  0.34123699
## [10,] -0.8761098 -0.06120426 -0.33734481  1.07126141  0.60422233
## [11,] -1.0300754  2.05497936  0.71272611 -0.18642264  1.36875777
## [12,] -1.7634714  2.67003031  1.02592796 -0.06283341 -1.24224080
## [13,] -0.5237739 -0.03817130  0.24834157  1.66783716 -0.01851435
## [14,] -0.3970150 -2.62600786  2.38221075 -0.65952544 -0.78205788
## [15,]  4.5051909  2.07587258  1.16496386 -0.54046226  0.51827932
## [16,] -0.3245045 -0.08311863  0.06883471 -0.15944735  0.33948840
## [17,] -1.0067379 -0.19861999 -0.23621440  0.21250565  0.34113454
## [18,]  0.2527034  0.98180077 -2.63568590 -1.18638398  1.37657195

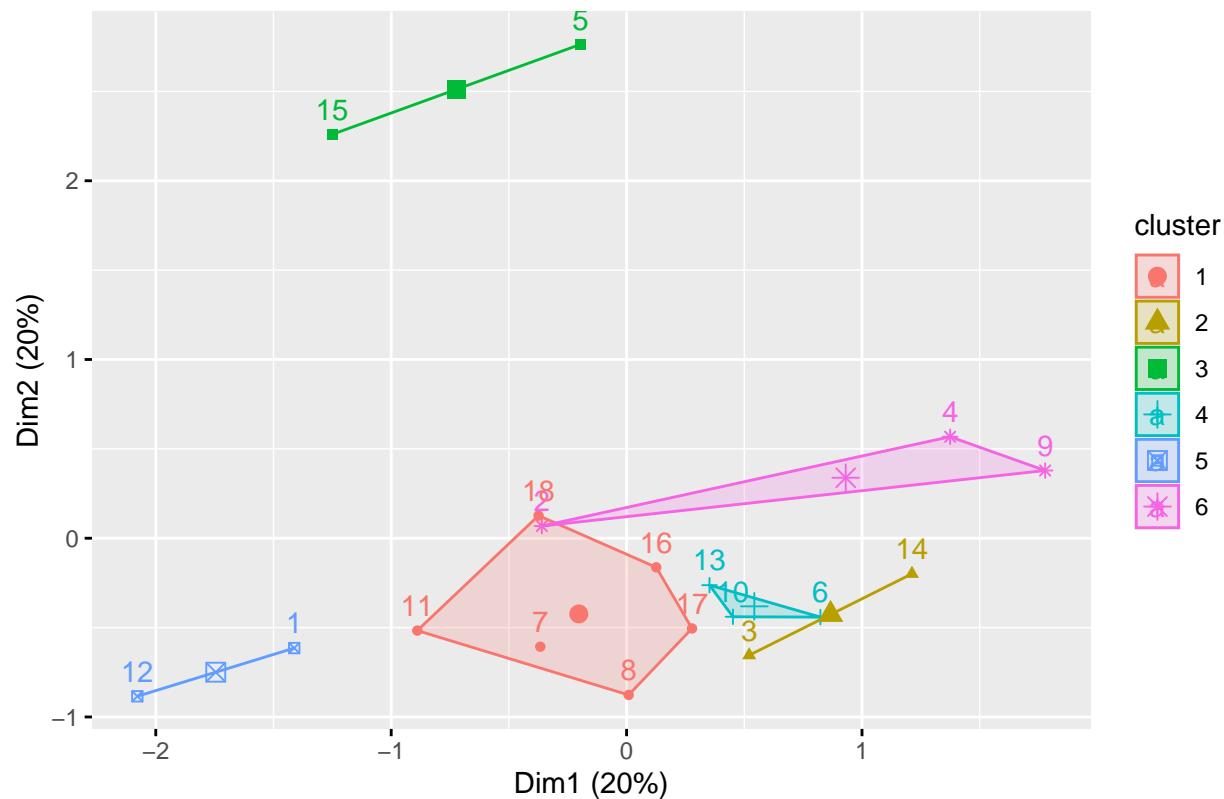
k = 6
kmeans_model = kmeans(X, centers = k, nstart=1)

waterQuality_2021_Avg$cluster=kmeans_model$cluster

fviz_cluster(kmeans_model, data = X)

```

Cluster plot



```
mean_by_group = aggregate(cbind(TemperatureAvg, DissolvedO2Avg, pHAvg, ConductivityAvg, BODAvg, NitrateAvg),
                          by = list(waterQuality_2021_Avg$State), FUN = mean)
table(waterQuality_2021_Avg$State, waterQuality_2021_Avg$cluster)
```

```
##
##                                     1 2 3 4 5 6
## ANDHRA PRADESH                  0 0 0 0 1 0
## DAMAN AND DIU, \r\nDADRA AND\r\nNAGAR HAVELI 0 0 0 0 0 1
## GOA                                0 1 0 0 0 0
## GUJARAT                            0 0 0 0 0 1
## HARYANA                            0 0 1 0 0 0
## HIMACHAL\r\nPRADESH                0 0 0 1 0 0
## KARNATAKA                           1 0 0 0 0 0
## KERALA                             1 0 0 0 0 0
## MAHARASHTRA                         0 0 0 0 0 1
## MANIPUR                            0 0 0 1 0 0
## MEGHALAYA                           1 0 0 0 0 0
## MIZORAM                            0 0 0 0 1 0
## ODISHA                             0 0 0 1 0 0
## PUDUCHERRY                         0 1 0 0 0 0
## PUNJAB                             0 0 1 0 0 0
## State Name                          1 0 0 0 0 0
## TAMIL NADU                          1 0 0 0 0 0
## TRIPURA                            1 0 0 0 0 0
```

waterQuality_2021_Avg

```
##                                     State TemperatureAvg DissolvedO2Avg
## 1          ANDHRA PRADESH      23.50000   6.983333
## 2  DAMAN AND DIU,\r\nDADRA AND\r\nNAGAR HAVELI      26.11000   5.940000
## 3                               GOA      27.17308   6.292308
## 4                               GUJARAT     24.29412   5.002941
## 5                               HARYANA     21.38959   5.094444
## 6  HIMACHAL\r\nPRADESH      17.33333   7.883333
## 7          KARNATAKA      25.00000   7.416667
## 8          KERALA      25.36620   6.465493
## 9          MAHARASHTRA     24.99701   6.506061
## 10         MANIPUR      19.08333   6.937500
## 11         MEGHALAYA     14.55833   6.345833
## 12         MIZORAM      16.61538   6.734615
## 13          ODISHA      15.50000   7.378571
## 14        PUDUCHERRY     26.80000   6.060000
## 15          PUNJAB      13.95294   3.532353
## 16          State Name     21.90126   6.294272
## 17          TAMIL NADU     24.50000   6.957143
## 18          TRIPURA      18.53125   5.690625
##          pHAvg ConductivityAvg    BODAvg NitrateAvg TotalColiformAvg
## 1  7.708333      366.1667  1.783333  1.7633333      150.2500
## 2  7.820000      5236.6500  4.065000  2.3830000     1827.7500
## 3  7.023077     14549.1923  1.671154  0.5980769     3225.5577
## 4  7.750000      7829.2353  4.994118  0.6073529     1097.5000
## 5  7.822222      1273.0556  20.177778  1.3876593    406828.5000
## 6  7.575000      726.3333  2.091667  0.8416667     14344.6667
## 7  7.175000      121.5833  2.250000  1.9233333      637.5000
## 8  6.942958      3742.0000  1.700704  0.5437324     2565.9155
## 9  7.534848      3783.1818  4.351515  1.4557576     1550.6515
## 10 7.269444      260.6250  3.397222  1.0101272      221.8750
## 11 6.922917      146.6042  6.875000  1.0356250    12142.9167
## 12 7.603846      213.9231  1.546154  1.0138462    17167.3474
## 13 7.742857      3743.0714  1.425000  0.8921429     2275.8929
## 14 7.670000     15654.2000  9.830000  0.3800000     34615.2025
## 15 7.514706      764.7941  21.176471  1.5891176    532942.0588
## 16 7.296440      3526.7307  4.499845  1.0513112    43034.0032
## 17 7.271429      285.9643  2.617857  0.8671429     159.3929
## 18 7.153125      172.5938  1.787500  2.2262500     237.6875
##          FecalColiformAvg      NO2      PM10      S02      PM2.5 cluster
## 1           12.83333  17.562500  28.25000  7.125000  17.56250      5
## 2           1128.35000 17.758012  51.72925  8.230870  26.77126      6
## 3           1854.50000 13.411765  59.06424  7.021524  21.26011      2
## 4           789.00000 24.562963  67.99259  16.120823  29.08395      6
## 5           253693.94444 24.576923  78.03846  12.346154  26.32933      3
## 6           10861.25000 11.119048  71.55952  2.226190  36.89954      4
## 7            556.83333 17.102624  47.58049  7.824026  24.67611      1
## 8           1392.42254 12.891386  43.55600  4.900582  24.99776      1
## 9            284.66667 34.694938  67.94619  11.944126  35.15770      6
## 10          72.31944 24.231229  38.50000  7.254072  32.64061      4
## 11          3798.75000  9.928571  39.37500  6.562912  23.85490      1
## 12          4610.55843  5.071429  29.77914  2.571429  14.86606      5
```

```

## 13      961.10714 17.290323 63.40261  7.091398 27.53333     4
## 14      9147.68153 11.500000 41.83333  6.500000 39.28123     2
## 15      54922.88235 17.832208 67.48132  7.546342 23.52170     3
## 16      11396.85191 17.758012 51.72925  8.230870 26.77126     1
## 17      50.07143 17.758012 51.72925  8.230870 26.77126     1
## 18      55.50000 13.000000 55.75000 21.000000 24.25000     1

Result_2021<- mean_by_group[order(mean_by_group$TotalColiformAvg, mean_by_group$BODAvg, mean_by_group$DissolvedO2Avg),]

Result_2021$Rank=c("A", "B", "C", "D", "E", "F")

class(Result_2021$cluster)

## [1] "integer"

class(waterQuality_2021_Avg$cluster)

## [1] "integer"

WaterAir2021 = merge(waterQuality_2021_Avg, Result_2021[,c("cluster","Rank")], by = "cluster", all.x = TRUE)

WaterQualityAlone = waterQuality_2021_Avg[, c("TemperatureAvg"
,"DissolvedO2Avg"
,"pHAvg"
,"ConductivityAvg"
,"BODAvg"
,"FecalColiformAvg"
,"NitrateAvg"
,"TotalColiformAvg"
)] 

# Center and scale the data
WaterQualityAlone_scaled = scale(WaterQualityAlone)

# Perform PCA
pca_result = princomp(WaterQualityAlone_scaled, center = TRUE, scale. = TRUE)

## Warning: In princomp.default(WaterQualityAlone_scaled, center = TRUE, scale. = TRUE) :
##   extra arguments 'center', 'scale.' will be disregarded

# Print the summary of PCA results
summary(pca_result)

## Importance of components:
##                               Comp.1    Comp.2    Comp.3    Comp.4    Comp.5
## Standard deviation      1.8028738 1.2809666 0.9985130 0.8835962 0.74712918
## Proportion of Variance 0.4301939 0.2171747 0.1319596 0.1033335 0.07387968
## Cumulative Proportion  0.4301939 0.6473686 0.7793282 0.8826618 0.95654145
##                               Comp.6    Comp.7    Comp.8
## Standard deviation      0.40284370 0.34949074 0.209586853
## Proportion of Variance 0.02147864 0.01616609 0.005813821
## Cumulative Proportion  0.97802009 0.99418618 1.0000000000

```

```

#reverse the signs
pca_result$rotation = -1*pca_result$rotation

#display principal components
pca_result$rotation

## numeric(0)

var_explained = pca_result$sdev^2 / sum(pca_result$sdev^2)

print(var_explained)

##      Comp.1      Comp.2      Comp.3      Comp.4      Comp.5      Comp.6
## 0.430193912 0.217174699 0.131959623 0.103333540 0.073879678 0.021478639
##      Comp.7      Comp.8
## 0.016166088 0.005813821

pc.comp = pca_result$scores

#86%
pc.comp1 = -1*pc.comp[,1]
pc.comp2 = -1*pc.comp[,2]
pc.comp3 = -1*pc.comp[,3]
pc.comp4 = -1*pc.comp[,4]

library(factoextra)
library(tidyr)
library(ggplot2)

X = cbind(pc.comp1, pc.comp2, pc.comp3, pc.comp4)

X

##      pc.comp1      pc.comp2      pc.comp3      pc.comp4
## [1,] -0.636169984  0.74640330 -1.52009222  0.23349884
## [2,]  0.008630588 -0.14172969 -2.51910497 -0.74004719
## [3,] -1.507822991 -2.42625519  0.86949418 -0.92580545
## [4,]  0.047543779 -1.73604674 -0.19928068  0.34068076
## [5,]  5.009327289 -0.56402427 -0.55702984  0.71219905
## [6,] -0.897242173  0.96542735  0.26189448  1.60434026
## [7,] -1.158219322  1.02250860 -0.77648863 -0.97004579
## [8,] -1.358170649 -0.59850240  1.21727004 -0.74128063
## [9,] -0.532837835 -0.22189757 -0.88138808 -0.23232190
## [10,] -0.762517915  0.87664101  0.63350467  0.25805932
## [11,] -0.197234860  1.37099510  1.85912512 -0.39913068
## [12,] -0.411738613  1.02094634  0.17450769  1.09247423
## [13,] -0.717672270  0.62082816  0.07633468  1.80357220
## [14,] -0.160911080 -3.20598728  0.02287780  0.51602614
## [15,]  4.877243826  0.53139417  1.02721699 -0.83171684
## [16,] -0.331298567 -0.07550033  0.37451155 -0.21896807
## [17,] -1.094635812  0.17740950  0.25283244 -0.02728996
## [18,] -0.176273410  1.63738994 -0.31618522 -1.47424428

```

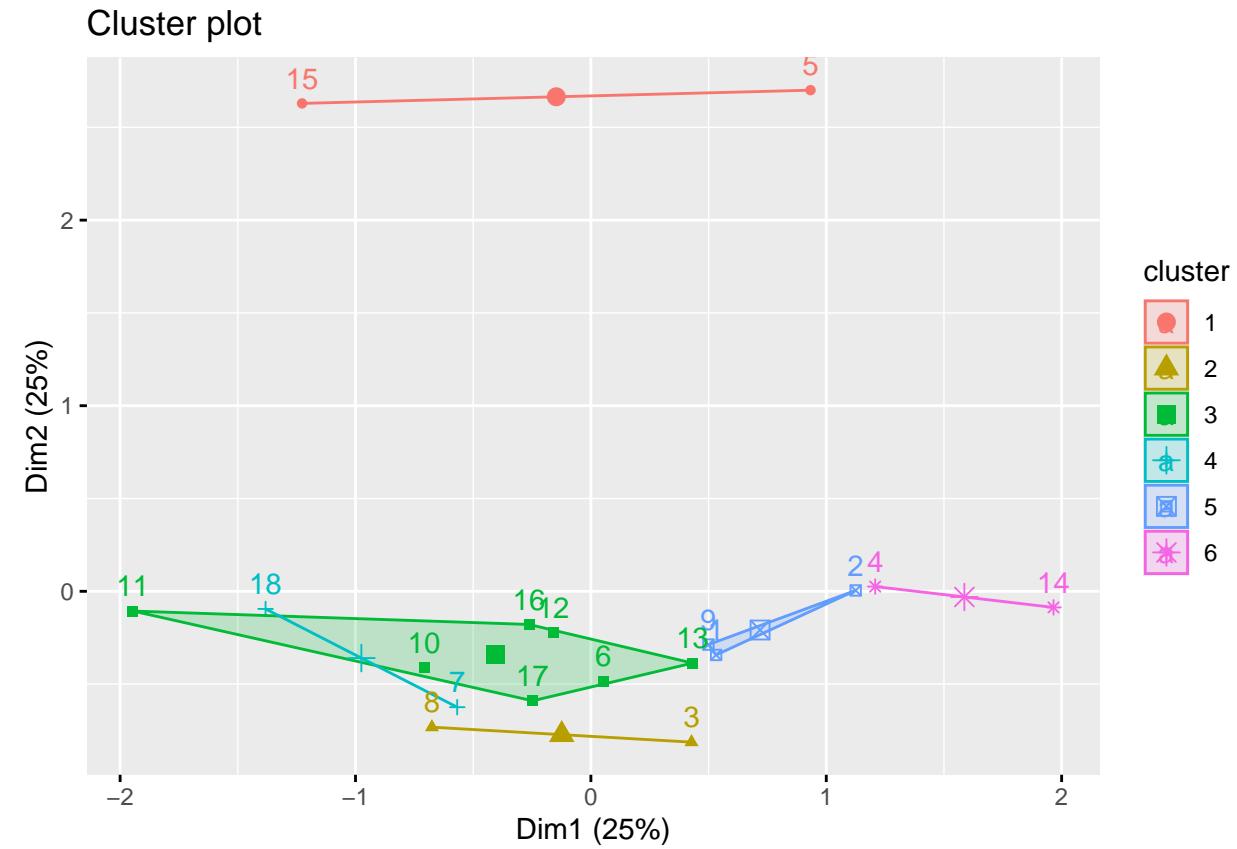
```

k = 6
kmeans_model = kmeans(X, centers = k, nstart=1)

waterQuality_2021_Avg$cluster=kmeans_model$cluster

fviz_cluster(kmeans_model, data = X)

```



```

mean_by_group = aggregate(cbind(TemperatureAvg, DissolvedO2Avg, pHAvg, ConductivityAvg, BODAvg, NitrateAvg),
                          by = list(waterQuality_2021_Avg$State))

table(waterQuality_2021_Avg$State, waterQuality_2021_Avg$cluster)

```

```

##                                     1 2 3 4 5 6
## ANDHRA PRADESH                   0 0 0 0 1 0
## DAMAN AND DIU, \r\nDADRA AND\r\nNAGAR HAVELI 0 0 0 0 1 0
## GOA                           0 1 0 0 0 0
## GUJARAT                      0 0 0 0 0 1
## HARYANA                       1 0 0 0 0 0
## HIMACHAL\r\nPRADESH            0 0 1 0 0 0
## KARNATAKA                     0 0 0 1 0 0
## KERALA                         0 1 0 0 0 0
## MAHARASHTRA                    0 0 0 0 1 0
## MANIPUR                        0 0 1 0 0 0

```

```

##    MEGHALAYA          0 0 1 0 0 0
##    MIZORAM            0 0 1 0 0 0
##    ODISHA              0 0 1 0 0 0
##    PUDUCHERRY         0 0 0 0 0 1
##    PUNJAB              1 0 0 0 0 0
##    State Name          0 0 1 0 0 0
##    TAMIL NADU          0 0 1 0 0 0
##    TRIPURA             0 0 0 1 0 0

```

WaterQualityAlone

```

##    TemperatureAvg DissolvedO2Avg     pHAvg ConductivityAvg      BODAvg
## 1      23.50000   6.983333 7.708333       366.1667 1.783333
## 2      26.11000   5.940000 7.820000       5236.6500 4.065000
## 3      27.17308   6.292308 7.023077      14549.1923 1.671154
## 4      24.29412   5.002941 7.750000      7829.2353 4.994118
## 5      21.38959   5.094444 7.822222      1273.0556 20.177778
## 6      17.33333   7.883333 7.575000      726.3333 2.091667
## 7      25.00000   7.416667 7.175000      121.5833 2.250000
## 8      25.36620   6.465493 6.942958      3742.0000 1.700704
## 9      24.99701   6.506061 7.534848      3783.1818 4.351515
## 10     19.08333   6.937500 7.269444      260.6250 3.397222
## 11     14.55833   6.345833 6.922917      146.6042 6.875000
## 12     16.61538   6.734615 7.603846      213.9231 1.546154
## 13     15.50000   7.378571 7.742857      3743.0714 1.425000
## 14     26.80000   6.060000 7.670000      15654.2000 9.830000
## 15     13.95294   3.532353 7.514706      764.7941 21.176471
## 16     21.90126   6.294272 7.296440      3526.7307 4.499845
## 17     24.50000   6.957143 7.271429      285.9643 2.617857
## 18     18.53125   5.690625 7.153125      172.5938 1.787500
##    FecalColiformAvg NitrateAvg TotalColiformAvg
## 1      12.83333   1.7633333 150.2500
## 2      1128.35000  2.3830000 1827.7500
## 3      1854.50000  0.5980769 3225.5577
## 4      789.00000  0.6073529 1097.5000
## 5      253693.94444 1.3876593 406828.5000
## 6      10861.25000 0.8416667 14344.6667
## 7      556.83333  1.9233333 637.5000
## 8      1392.42254  0.5437324 2565.9155
## 9      284.66667  1.4557576 1550.6515
## 10     72.31944  1.0101272 221.8750
## 11     3798.75000 1.0356250 12142.9167
## 12     4610.55843 1.0138462 17167.3474
## 13     961.10714  0.8921429 2275.8929
## 14     9147.68153 0.3800000 34615.2025
## 15     54922.88235 1.5891176 532942.0588
## 16     11396.85191 1.0513112 43034.0032
## 17      50.07143  0.8671429 159.3929
## 18      55.50000  2.2262500 237.6875

```

```
Result_2021_water= mean_by_group[order(mean_by_group$TotalColiformAvg, mean_by_group$BODAvg, mean_by_gr
```

```

Result_2021_water$Rank=c("A", "B", "C", "D", "E", "F")
class(waterQuality_2021_Avg$cluster)

## [1] "integer"

class(waterQuality_2021_Avg$cluster)

## [1] "integer"

Water2021 = merge(waterQuality_2021_Avg, Result_2021_water[,c("cluster","Rank")], by = "cluster", all.x = TRUE)

library(stringr)
library(ggplot2)
library(sf)

## Linking to GEOS 3.11.1, GDAL 3.6.2, PROJ 9.1.1; sf_use_s2() is TRUE

plotD = WaterAir2017

# create a list of selected states with ranks
plotD$State[plotD$State == "DAMAN AND DIU, DADRA AND NAGAR HAVELI"] = "DAMAN AND DIU"

selected_states = str_to_title(plotD$State)

ranks = plotD$Rank
state_ranks = data.frame(state = selected_states, rank = ranks)

# read in the shapefile data for Indian states
india = st_read("C:/Users/SRINU/Desktop/DPA Project/IND_adm1.shp")

## Reading layer 'IND_adm1' from data source
##   'C:\Users\SRINU\Desktop\Project\IND_adm1.shp' using driver 'ESRI Shapefile'
## Simple feature collection with 36 features and 9 fields
## Geometry type: MULTIPOLYGON
## Dimension:      XY
## Bounding box:  xmin: 68.18625 ymin: 6.754256 xmax: 97.41516 ymax: 35.50133
## Geodetic CRS:  WGS 84

# join the state rank data with the shapefile data
india_ranks = merge(india, state_ranks, by.x = "NAME_1", by.y = "state")

# manually specify the colors for each rank
colors = c("#8BC34A", "#FFEB3B", "#9C27B0", "#673AB7", "#FF9800", "#F44336")
names(colors) = c("A", "B", "C", "D", "E", "F")

states <- india %>%
  st_centroid() %>%

```

```

st_coordinates() %>%
as.data.frame() %>%
mutate(state = india$NAME_1)

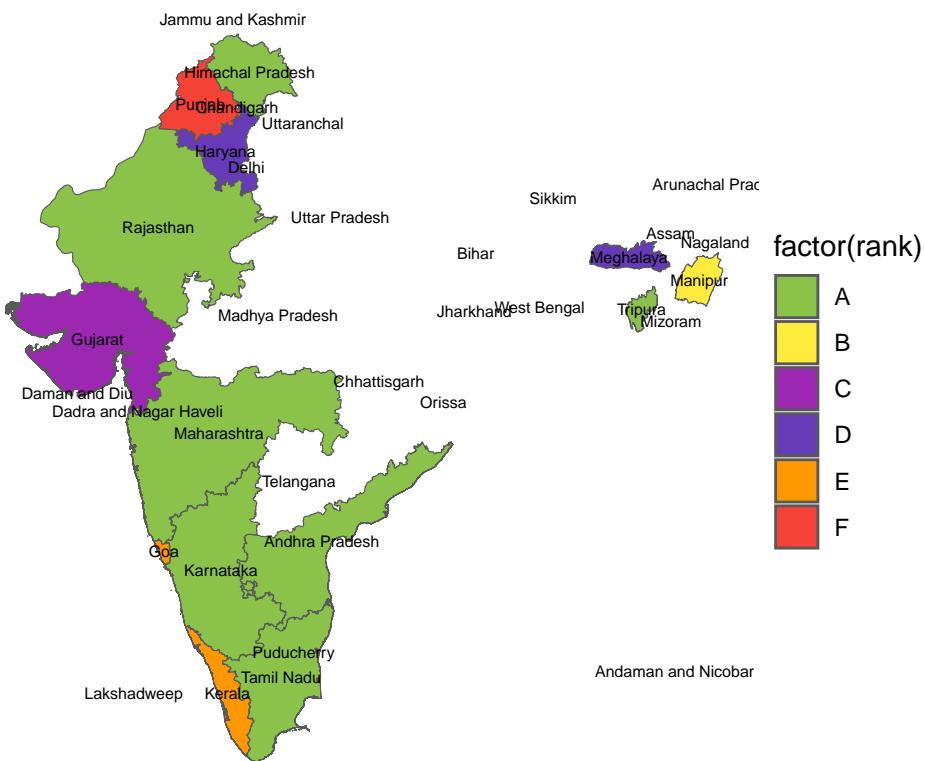
## Warning: st_centroid assumes attributes are constant over geometries

ggplot(data = india_ranks) +
  geom_sf(aes(fill = factor(rank))) +
  scale_fill_manual(values = colors) +
  theme_void() +
  geom_text(data = states, aes(x = X, y = Y, label = state), size = 2, color = "black", text = element_text())
  labs(title = "State-wise Rankings of Water and Air Quality in India 2017")

## Warning in geom_text(data = states, aes(x = X, y = Y, label = state), size = 2,
## : Ignoring unknown parameters: 'text'

```

State-wise Rankings of Water and Air Quality in India 2017



```

library(stringr)
library(ggplot2)
library(sf)

ploDWater = WaterAir2017

# create a list of selected states with ranks

```

```

ploDWater$State[ploDWater$State == "DAMAN AND DIU, DADRA AND NAGAR HAVELI"] = "DAMAN AND DIU"

selected_states = str_to_title(ploDWater$State)

ranks = ploDWater$Rank
state_ranks = data.frame(state = selected_states, rank = ranks)

# read in the shapefile data for Indian states
india = st_read("C:/Users/SRINU/Desktop/DPA Project/IND_adm1.shp")

## Reading layer 'IND_adm1' from data source
##   'C:\Users\SRINU\Desktop\Project\IND_adm1.shp' using driver 'ESRI Shapefile'
## Simple feature collection with 36 features and 9 fields
## Geometry type: MULTIPOLYGON
## Dimension:      XY
## Bounding box:  xmin: 68.18625 ymin: 6.754256 xmax: 97.41516 ymax: 35.50133
## Geodetic CRS:  WGS 84

# join the state rank data with the shapefile data
india_ranks = merge(india, state_ranks, by.x = "NAME_1", by.y = "state")

# manually specify the colors for each rank
colors = c("#8BC34A", "#FFEB3B", "#9C27B0", "#673AB7", "#FF9800", "#F44336")
names(colors) = c("A", "B", "C", "D", "E", "F")

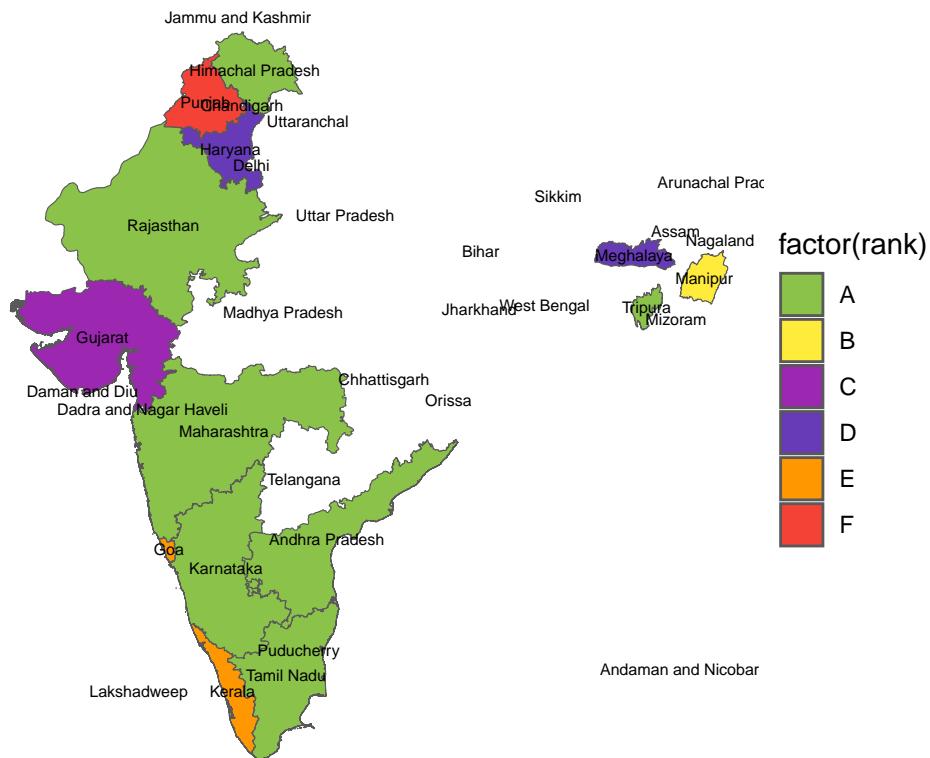
states <- india %>%
  st_centroid() %>%
  st_coordinates() %>%
  as.data.frame() %>%
  mutate(state = india$NAME_1)

## Warning: st_centroid assumes attributes are constant over geometries

ggplot(data = india_ranks) +
  geom_sf(aes(fill = factor(rank))) +
  scale_fill_manual(values = colors) +
  theme_void() +
  geom_text(data = states, aes(x = X, y = Y, label = state), size = 2, color = "black", text = element_text(vjust = 0))
  labs(title = "State-wise Rankings of Water Quality in India 2017")

## Warning in geom_text(data = states, aes(x = X, y = Y, label = state), size = 2,
## : Ignoring unknown parameters: 'text'
```

State-wise Rankings of Water Quality in India 2017



```

library(stringr)
library(ggplot2)
library(sf)

plotD = WaterAir2018

# create a list of selected states with ranks
plotD$State[plotD$State == "DAMAN AND DIU, DADRA AND NAGAR HAVELI"] = "DAMAN AND DIU"

selected_states = str_to_title(plotD$State)

ranks = plotD$Rank
state_ranks = data.frame(state = selected_states, rank = ranks)

# read in the shapefile data for Indian states
india = st_read("C:/Users/SRINU/Desktop/DPA Project/IND_adm1.shp")

## Reading layer 'IND_adm1' from data source
##   'C:\Users\SRINU\Desktop\DP Project\IND_adm1.shp' using driver 'ESRI Shapefile'
## Simple feature collection with 36 features and 9 fields
## Geometry type: MULTIPOLYGON
## Dimension:      XY
## Bounding box:  xmin: 68.18625 ymin: 6.754256 xmax: 97.41516 ymax: 35.50133
## Geodetic CRS:  WGS 84

```

```

# join the state rank data with the shapefile data
india_ranks = merge(india, state_ranks, by.x = "NAME_1", by.y = "state")

# manually specify the colors for each rank
colors = c("#8BC34A", "#FFEB3B", "#9C27B0", "#673AB7", "#FF9800", "#F44336")
names(colors) = c("A", "B", "C", "D", "E", "F")

states <- india %>%
  st_centroid() %>%
  st_coordinates() %>%
  as.data.frame() %>%
  mutate(state = india$NAME_1)

```

Warning: st_centroid assumes attributes are constant over geometries

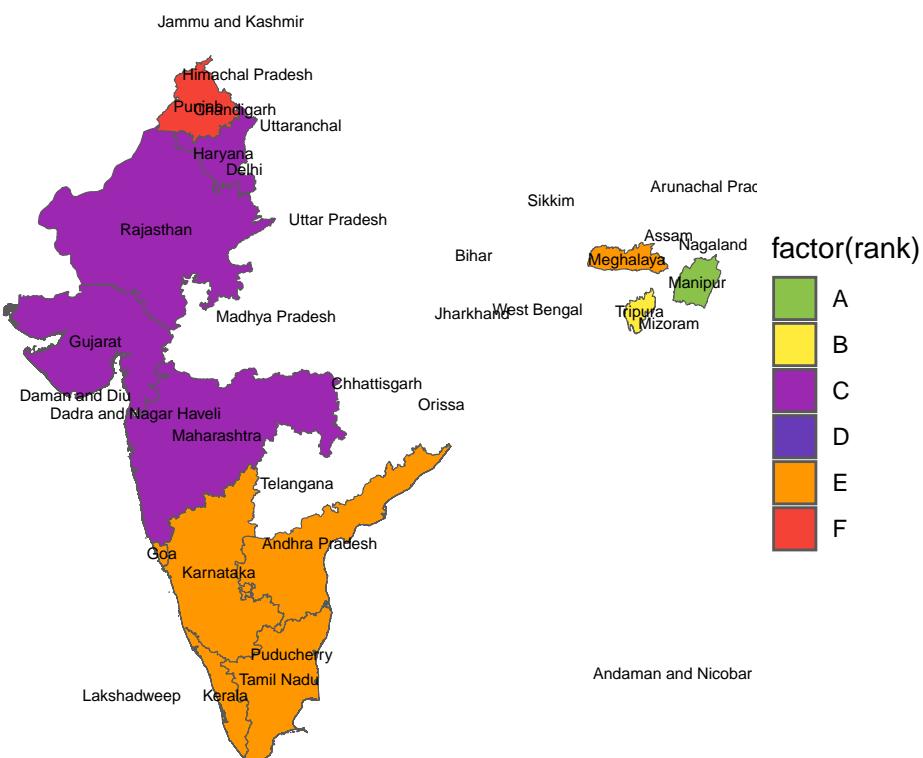
```

ggplot(data = india_ranks) +
  geom_sf(aes(fill = factor(rank))) +
  scale_fill_manual(values = colors) +
  theme_void() +
  geom_text(data = states, aes(x = X, y = Y, label = state), size = 2, color = "black", text = element_text())
  labs(title = "State-wise Rankings of Water and Air Quality in India 2018")

```

Warning in geom_text(data = states, aes(x = X, y = Y, label = state), size = 2,
: Ignoring unknown parameters: 'text'

State-wise Rankings of Water and Air Quality in India 2018



```

library(stringr)
library(ggplot2)
library(sf)

ploDWater = Water2018

# create a list of selected states with ranks
ploDWater$State[ploDWater$State == "DAMAN AND DIU, DADRA AND NAGAR HAVELI"] = "DAMAN AND DIU"

selected_states = str_to_title(ploDWater$State)

ranks = ploDWater$Rank
state_ranks = data.frame(state = selected_states, rank = ranks)

# read in the shapefile data for Indian states
india = st_read("C:/Users/SRINU/Desktop/DPA Project/IND_adm1.shp")

## Reading layer 'IND_adm1' from data source
##   'C:\Users\SRINU\Desktop\Project\IND_adm1.shp' using driver 'ESRI Shapefile'
## Simple feature collection with 36 features and 9 fields
## Geometry type: MULTIPOLYGON
## Dimension:      XY
## Bounding box:  xmin: 68.18625 ymin: 6.754256 xmax: 97.41516 ymax: 35.50133
## Geodetic CRS:  WGS 84

# join the state rank data with the shapefile data
india_ranks = merge(india, state_ranks, by.x = "NAME_1", by.y = "state")

# manually specify the colors for each rank
colors = c("#8BC34A", "#FFEB3B", "#9C27B0", "#673AB7", "#FF9800", "#F44336")
names(colors) = c("A", "B", "C", "D", "E", "F")

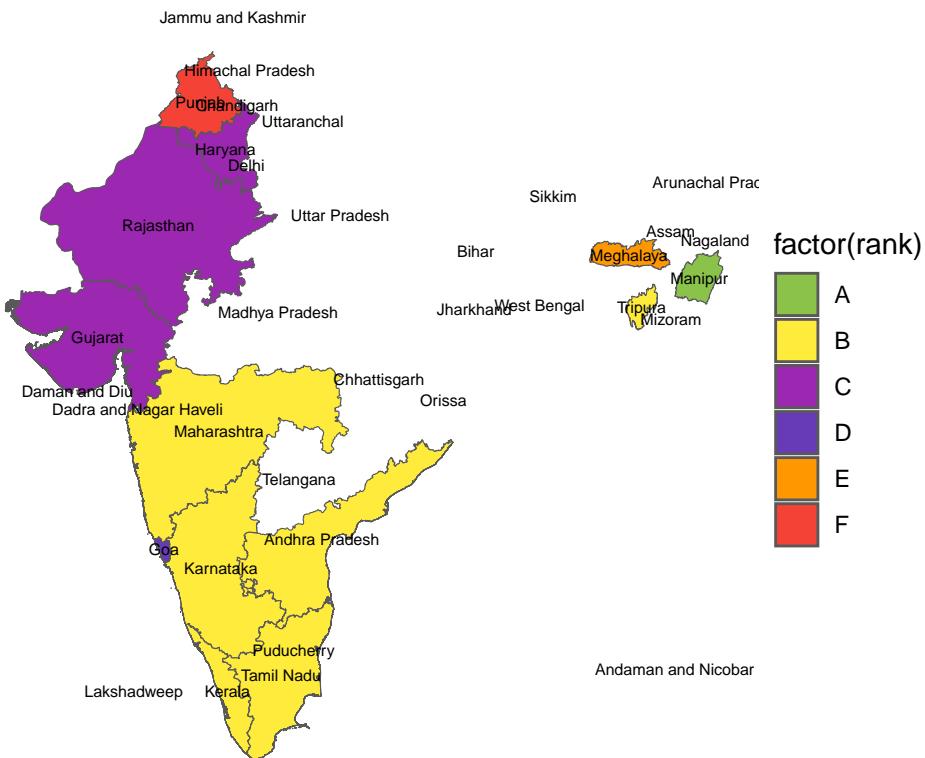
states <- india %>%
  st_centroid() %>%
  st_coordinates() %>%
  as.data.frame() %>%
  mutate(state = india$NAME_1)

## Warning: st_centroid assumes attributes are constant over geometries

ggplot(data = india_ranks) +
  geom_sf(aes(fill = factor(rank))) +
  scale_fill_manual(values = colors) +
  theme_void() +
  geom_text(data = states, aes(x = X, y = Y, label = state), size = 2, color = "black", text = element_text(vjust = 0))
  labs(title = "State-wise Rankings of Water Quality in India 2018")

## Warning in geom_text(data = states, aes(x = X, y = Y, label = state), size = 2,
## : Ignoring unknown parameters: 'text'
```

State-wise Rankings of Water Quality in India 2018



```

library(stringr)
library(ggplot2)
library(sf)

plotD = WaterAir2019

# create a list of selected states with ranks
plotD$State[plotD$State == "DAMAN AND DIU, DADRA AND NAGAR HAVELI"] = "DAMAN AND DIU"

selected_states = str_to_title(plotD$State)

ranks = plotD$Rank
state_ranks = data.frame(state = selected_states, rank = ranks)

# read in the shapefile data for Indian states
india = st_read("C:/Users/SRINU/Desktop/DPA Project/IND_adm1.shp")

## Reading layer 'IND_adm1' from data source
##   'C:\Users\SRINU\Desktop\DP Project\IND_adm1.shp' using driver 'ESRI Shapefile'
## Simple feature collection with 36 features and 9 fields
## Geometry type: MULTIPOLYGON
## Dimension:      XY
## Bounding box:  xmin: 68.18625 ymin: 6.754256 xmax: 97.41516 ymax: 35.50133
## Geodetic CRS:  WGS 84

```

```

# join the state rank data with the shapefile data
india_ranks = merge(india, state_ranks, by.x = "NAME_1", by.y = "state")

# manually specify the colors for each rank
colors = c("#8BC34A", "#FFEB3B", "#9C27B0", "#673AB7", "#FF9800", "#F44336")
names(colors) = c("A", "B", "C", "D", "E", "F")

states <- india %>%
  st_centroid() %>%
  st_coordinates() %>%
  as.data.frame() %>%
  mutate(state = india$NAME_1)

```

Warning: st_centroid assumes attributes are constant over geometries

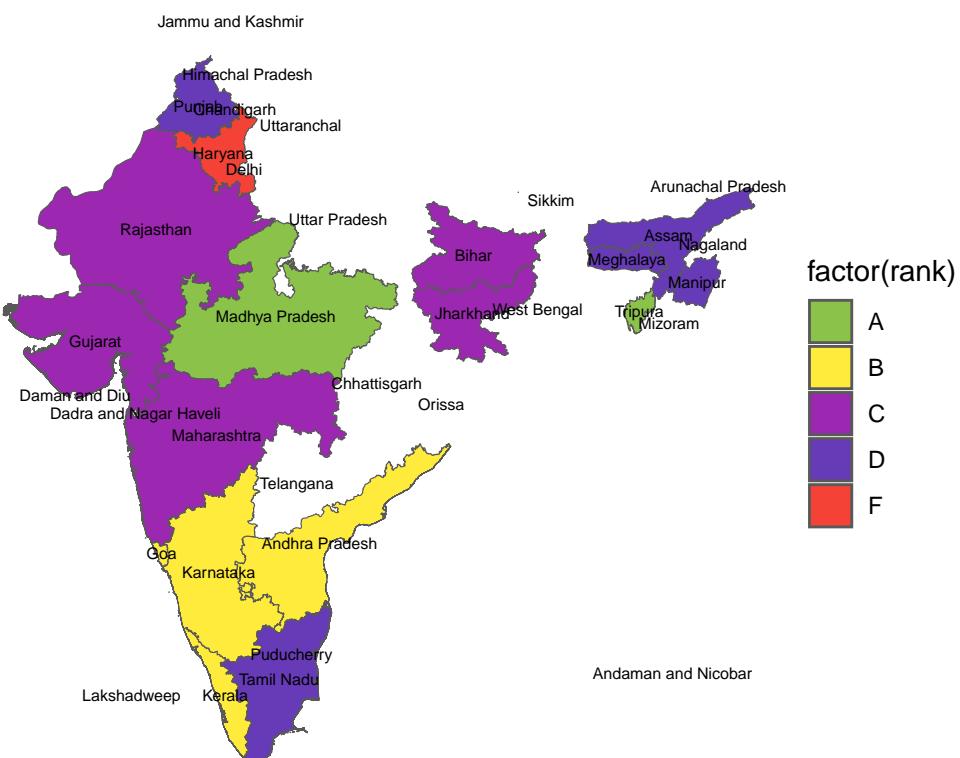
```

ggplot(data = india_ranks) +
  geom_sf(aes(fill = factor(rank))) +
  scale_fill_manual(values = colors) +
  theme_void() +
  geom_text(data = states, aes(x = X, y = Y, label = state), size = 2, color = "black", text = element_text())
  labs(title = "State-wise Rankings of Water and Air Quality in India 2019")

```

Warning in geom_text(data = states, aes(x = X, y = Y, label = state), size = 2,
: Ignoring unknown parameters: 'text'

State-wise Rankings of Water and Air Quality in India 2019



```

library(stringr)
library(ggplot2)
library(sf)

ploDWater = Water2019

# create a list of selected states with ranks
ploDWater$State[ploDWater$State == "DAMAN AND DIU, DADRA AND NAGAR HAVELI"] = "DAMAN AND DIU"

selected_states = str_to_title(ploDWater$State)

ranks = ploDWater$Rank
state_ranks = data.frame(state = selected_states, rank = ranks)

# read in the shapefile data for Indian states
india = st_read("C:/Users/SRINU/Desktop/DPA Project/IND_adm1.shp")

## Reading layer 'IND_adm1' from data source
##   'C:\Users\SRINU\Desktop\Project\IND_adm1.shp' using driver 'ESRI Shapefile'
## Simple feature collection with 36 features and 9 fields
## Geometry type: MULTIPOLYGON
## Dimension:      XY
## Bounding box:  xmin: 68.18625 ymin: 6.754256 xmax: 97.41516 ymax: 35.50133
## Geodetic CRS:  WGS 84

# join the state rank data with the shapefile data
india_ranks = merge(india, state_ranks, by.x = "NAME_1", by.y = "state")

# manually specify the colors for each rank
colors = c("#8BC34A", "#FFEB3B", "#9C27B0", "#673AB7", "#FF9800", "#F44336")
names(colors) = c("A", "B", "C", "D", "E", "F")

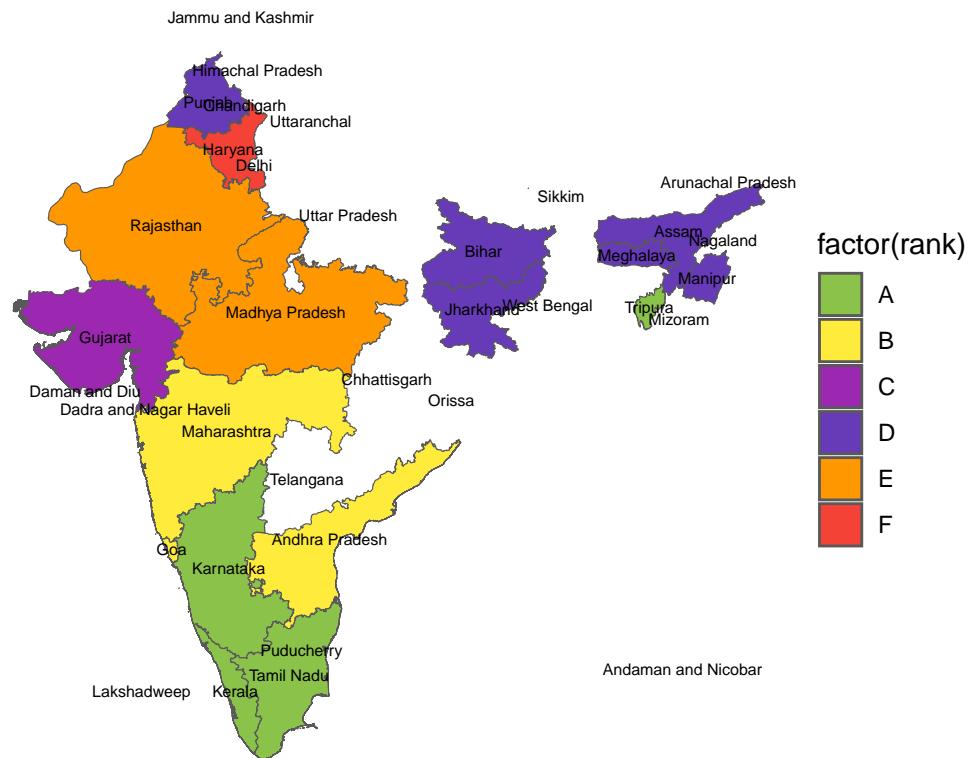
states <- india %>%
  st_centroid() %>%
  st_coordinates() %>%
  as.data.frame() %>%
  mutate(state = india$NAME_1)

## Warning: st_centroid assumes attributes are constant over geometries

ggplot(data = india_ranks) +
  geom_sf(aes(fill = factor(rank))) +
  scale_fill_manual(values = colors) +
  theme_void() +
  geom_text(data = states, aes(x = X, y = Y, label = state), size = 2, color = "black", text = element_text(vjust = 0))
  labs(title = "State-wise Rankings of Water Quality in India 2019")

## Warning in geom_text(data = states, aes(x = X, y = Y, label = state), size = 2,
## : Ignoring unknown parameters: 'text'
```

State-wise Rankings of Water Quality in India 2019



```

library(stringr)
library(ggplot2)
library(sf)

plotD = WaterAir2020

# create a list of selected states with ranks
plotD$State[plotD$State == "DAMAN AND DIU, DADRA AND NAGAR HAVELI"] = "DAMAN AND DIU"

selected_states = str_to_title(plotD$State)

ranks = plotD$Rank
state_ranks = data.frame(state = selected_states, rank = ranks)

# read in the shapefile data for Indian states
india = st_read("C:/Users/SRINU/Desktop/DPA Project/IND_adm1.shp")

## Reading layer 'IND_adm1' from data source
##   'C:\Users\SRINU\Desktop\DP Project\IND_adm1.shp' using driver 'ESRI Shapefile'
## Simple feature collection with 36 features and 9 fields
## Geometry type: MULTIPOLYGON
## Dimension:      XY
## Bounding box:  xmin: 68.18625 ymin: 6.754256 xmax: 97.41516 ymax: 35.50133
## Geodetic CRS:  WGS 84

```

```

# join the state rank data with the shapefile data
india_ranks = merge(india, state_ranks, by.x = "NAME_1", by.y = "state")

# manually specify the colors for each rank
colors = c("#8BC34A", "#FFEB3B", "#9C27B0", "#673AB7", "#FF9800", "#F44336")
names(colors) = c("A", "B", "C", "D", "E", "F")

states <- india %>%
  st_centroid() %>%
  st_coordinates() %>%
  as.data.frame() %>%
  mutate(state = india$NAME_1)

```

Warning: st_centroid assumes attributes are constant over geometries

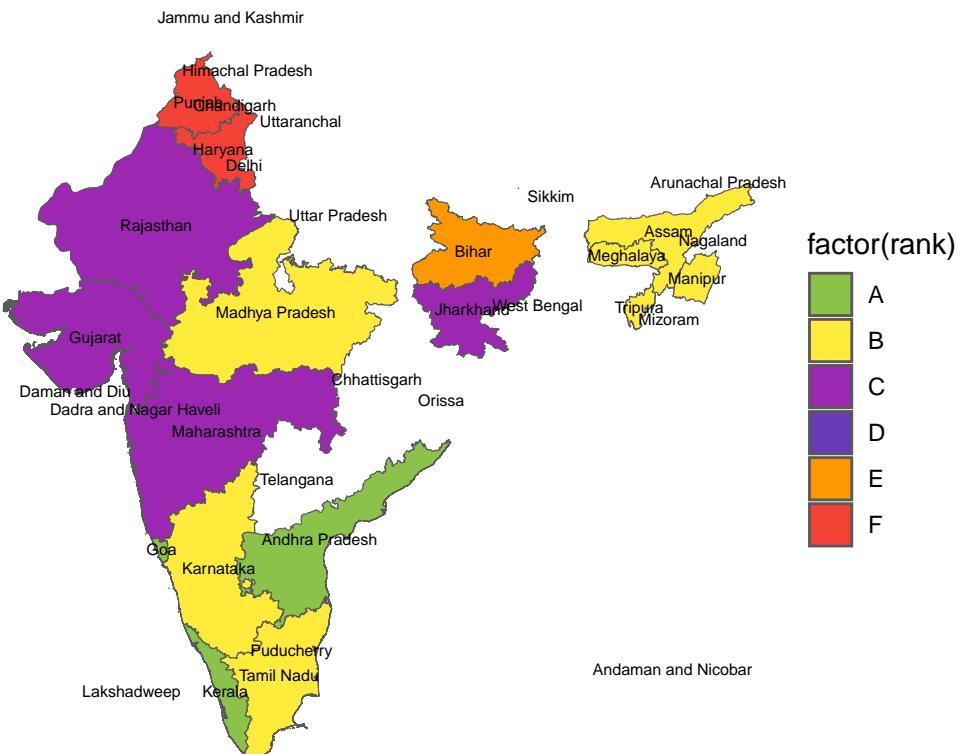
```

ggplot(data = india_ranks) +
  geom_sf(aes(fill = factor(rank))) +
  scale_fill_manual(values = colors) +
  theme_void() +
  geom_text(data = states, aes(x = X, y = Y, label = state), size = 2, color = "black", text = element_text())
  labs(title = "State-wise Rankings of Water and Air Quality in India 2020")

```

Warning in geom_text(data = states, aes(x = X, y = Y, label = state), size = 2,
: Ignoring unknown parameters: 'text'

State-wise Rankings of Water and Air Quality in India 2020



```

library(stringr)
library(ggplot2)
library(sf)

ploDWater = Water2020

# create a list of selected states with ranks
ploDWater$State[ploDWater$State == "DAMAN AND DIU, DADRA AND NAGAR HAVELI"] = "DAMAN AND DIU"

selected_states = str_to_title(ploDWater$State)

ranks = ploDWater$Rank
state_ranks = data.frame(state = selected_states, rank = ranks)

# read in the shapefile data for Indian states
india = st_read("C:/Users/SRINU/Desktop/DPA Project/IND_adm1.shp")

## Reading layer 'IND_adm1' from data source
##   'C:\Users\SRINU\Desktop\Project\IND_adm1.shp' using driver 'ESRI Shapefile'
## Simple feature collection with 36 features and 9 fields
## Geometry type: MULTIPOLYGON
## Dimension:      XY
## Bounding box:  xmin: 68.18625 ymin: 6.754256 xmax: 97.41516 ymax: 35.50133
## Geodetic CRS:  WGS 84

# join the state rank data with the shapefile data
india_ranks = merge(india, state_ranks, by.x = "NAME_1", by.y = "state")

# manually specify the colors for each rank
colors = c("#8BC34A", "#FFEB3B", "#9C27B0", "#673AB7", "#FF9800", "#F44336")
names(colors) = c("A", "B", "C", "D", "E", "F")

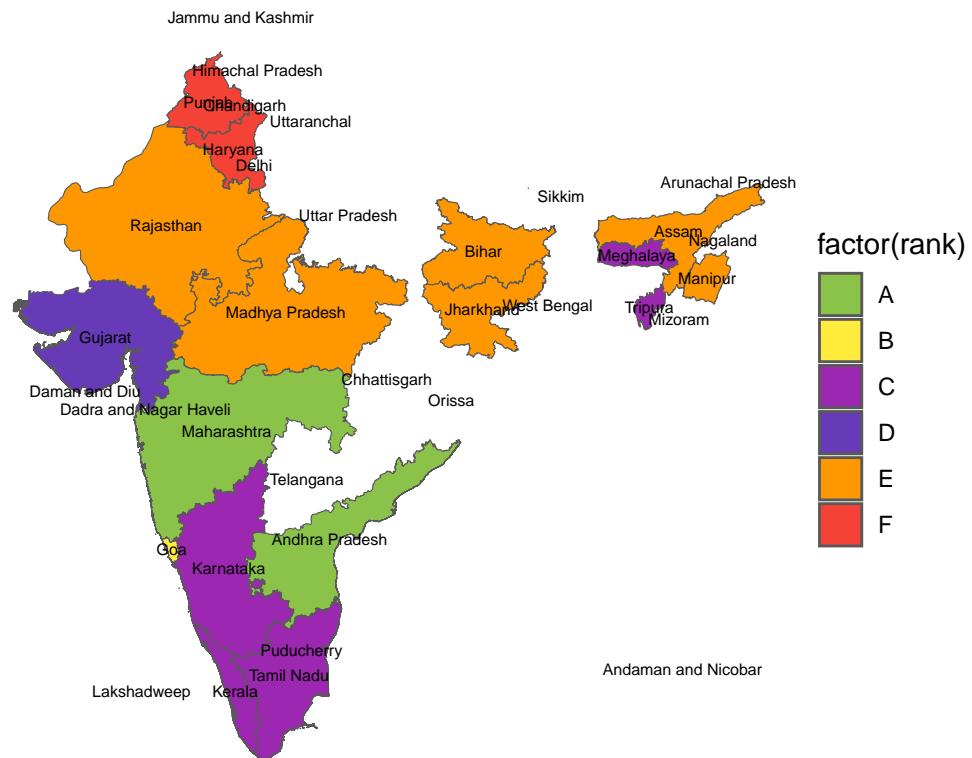
states <- india %>%
  st_centroid() %>%
  st_coordinates() %>%
  as.data.frame() %>%
  mutate(state = india$NAME_1)

## Warning: st_centroid assumes attributes are constant over geometries

ggplot(data = india_ranks) +
  geom_sf(aes(fill = factor(rank))) +
  scale_fill_manual(values = colors) +
  theme_void() +
  geom_text(data = states, aes(x = X, y = Y, label = state), size = 2, color = "black", text = element_text(vjust = 0))
  labs(title = "State-wise Rankings of Water Quality in India 2020")

## Warning in geom_text(data = states, aes(x = X, y = Y, label = state), size = 2,
## : Ignoring unknown parameters: 'text'
```

State-wise Rankings of Water Quality in India 2020



```

library(stringr)
library(ggplot2)
library(sf)

plotD = WaterAir2021

# create a list of selected states with ranks
plotD$State[plotD$State == "DAMAN AND DIU, DADRA AND NAGAR HAVELI"] = "DAMAN AND DIU"

selected_states = str_to_title(plotD$State)

ranks = plotD$Rank
state_ranks = data.frame(state = selected_states, rank = ranks)

# read in the shapefile data for Indian states
india = st_read("C:/Users/SRINU/Desktop/DPA Project/IND_adm1.shp")

## Reading layer 'IND_adm1' from data source
##   'C:\Users\SRINU\Desktop\Project\IND_adm1.shp' using driver 'ESRI Shapefile'
## Simple feature collection with 36 features and 9 fields
## Geometry type: MULTIPOLYGON
## Dimension:      XY
## Bounding box:  xmin: 68.18625 ymin: 6.754256 xmax: 97.41516 ymax: 35.50133
## Geodetic CRS:  WGS 84

```

```

# join the state rank data with the shapefile data
india_ranks = merge(india, state_ranks, by.x = "NAME_1", by.y = "state")

# manually specify the colors for each rank
colors = c("#8BC34A", "#FFEB3B", "#9C27B0", "#673AB7", "#FF9800", "#F44336")
names(colors) = c("A", "B", "C", "D", "E", "F")

states <- india %>%
  st_centroid() %>%
  st_coordinates() %>%
  as.data.frame() %>%
  mutate(state = india$NAME_1)

```

Warning: st_centroid assumes attributes are constant over geometries

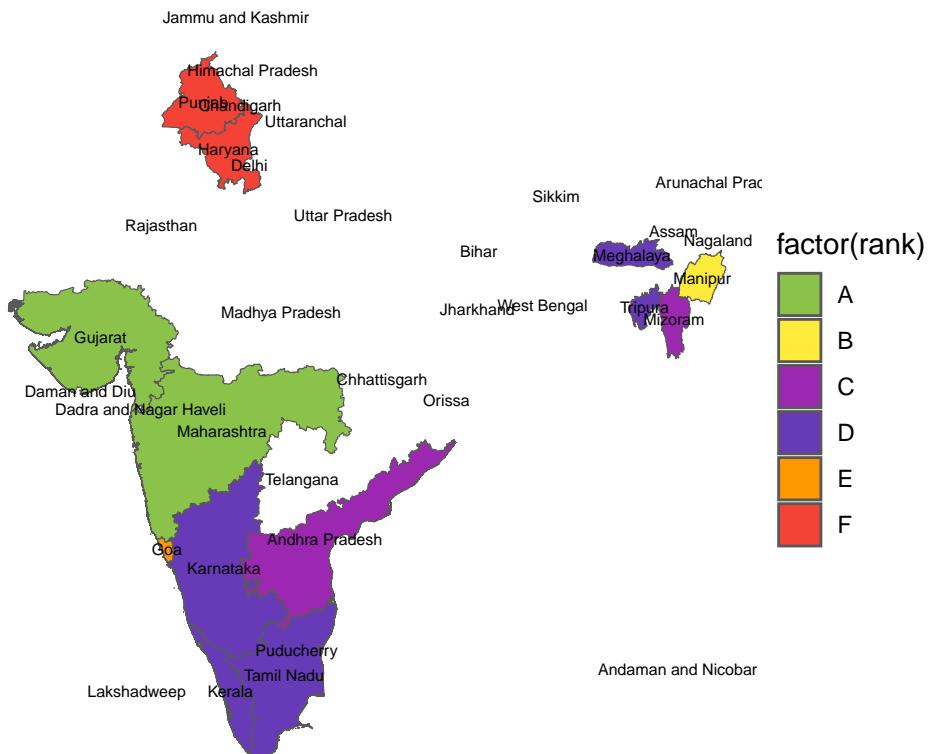
```

ggplot(data = india_ranks) +
  geom_sf(aes(fill = factor(rank))) +
  scale_fill_manual(values = colors) +
  theme_void() +
  geom_text(data = states, aes(x = X, y = Y, label = state), size = 2, color = "black", text = element_text())
  labs(title = "State-wise Rankings of Water and Air Quality in India 2021")

```

Warning in geom_text(data = states, aes(x = X, y = Y, label = state), size = 2,
: Ignoring unknown parameters: 'text'

State-wise Rankings of Water and Air Quality in India 2021



```

library(stringr)
library(ggplot2)
library(sf)

ploDWater = Water2021

# create a list of selected states with ranks
ploDWater$State[ploDWater$State == "DAMAN AND DIU, DADRA AND NAGAR HAVELI"] = "DAMAN AND DIU"

selected_states = str_to_title(ploDWater$State)

ranks = ploDWater$Rank
state_ranks = data.frame(state = selected_states, rank = ranks)

# read in the shapefile data for Indian states
india = st_read("C:/Users/SRINU/Desktop/DPA Project/IND_adm1.shp")

## Reading layer 'IND_adm1' from data source
##   'C:\Users\SRINU\Desktop\DPA Project\IND_adm1.shp' using driver 'ESRI Shapefile'
## Simple feature collection with 36 features and 9 fields
## Geometry type: MULTIPOLYGON
## Dimension:      XY
## Bounding box:  xmin: 68.18625 ymin: 6.754256 xmax: 97.41516 ymax: 35.50133
## Geodetic CRS:  WGS 84

# join the state rank data with the shapefile data
india_ranks = merge(india, state_ranks, by.x = "NAME_1", by.y = "state")

# manually specify the colors for each rank
colors = c("#8BC34A", "#FFEB3B", "#9C27B0", "#673AB7", "#FF9800", "#F44336")
names(colors) = c("A", "B", "C", "D", "E", "F")

states <- india %>%
  st_centroid() %>%
  st_coordinates() %>%
  as.data.frame() %>%
  mutate(state = india$NAME_1)

## Warning: st_centroid assumes attributes are constant over geometries

ggplot(data = india_ranks) +
  geom_sf(aes(fill = factor(rank))) +
  scale_fill_manual(values = colors) +
  theme_void() +
  geom_text(data = states, aes(x = X, y = Y, label = state), size = 2, color = "black", text = element_text(vjust = 0))
  labs(title = "State-wise Rankings of Water Quality in India 2021")

## Warning in geom_text(data = states, aes(x = X, y = Y, label = state), size = 2,
## : Ignoring unknown parameters: 'text'
```

State-wise Rankings of Water Quality in India 2021

