

# Stats525HW7

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2.4 a. Obtain a 99 percent confidence interval for  $b_1$ . Interpret your confidence interval. Does it include zero? Why might the director of admissions be interested in whether the confidence interval includes zero? CI: (0.00175, 0.094787) This means that we can be 95% confident that for every increase in GPA, the ACT score will increase by between 0.00175 and 0.094787. This does not include zero. The director would be interested in whether the confidence interval includes zero because if it did then there might not be a correlation between ACT scores and GPA.

```
library(tidyverse)
```

```
## — Attaching packages — tidyverse
1.3.0 —
```

```
## ✓ ggplot2 3.3.2      ✓ purrr  0.3.4
## ✓ tibble  3.0.3      ✓ dplyr  1.0.2
## ✓ tidyr   1.1.2      ✓ stringr 1.4.0
## ✓ readr   1.3.1      ✓ forcats 0.5.0
```

```
## — Conflicts — tidyverse_conflicts() —
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

```
loadRData <- function(fileName) {
  load(fileName)
  get(ls()[ls() != "fileName"])
}
gpa <- loadRData("/Users/lukegeel/Downloads/gpa_spring2021.RData")
n <- nrow(gpa)
y <- gpa$Y
x <- gpa$X
linmod <- lm(y~x)
b1 <- linmod$coef[2]
s.b1 <- summary(linmod)$coef[2, 2]
alpha <- 0.01
qt <- qt(alpha/2, n - 2)
pvalue <- pt(-abs(b1/s.b1), n-2) + (1-pt(abs(b1/s.b1), n-2))
lower <- b1+s.b1*qt(alpha/2, n-2)
upper <- b1-s.b1*qt(alpha/2, n-2)
lower
```

```
##           x
## 0.001748182
```

```
upper
```

```
##           x
## 0.09476612
```

- b. Test, using the test statistic  $t^*$ , whether or not a linear association exists between student's ACT score (X) and GPA at the end of the freshman year (Y). Use a level of significance of .01. State the alternatives, decision rule, and conclusion. Alternative hypothesis: There is no linear association between student's ACT score and GPA at the end of the freshman year. Decision rule: If the test statistic exceeds the t-value given the alpha level, we reject the alternative hypothesis. Conclusion: The test statistic exceeds t, so we can reject the alternative hypothesis and conclude that there is a linear association between ACT scores and GPA.

```
library(tidyverse)
loadRData <- function(fileName) {
  load(fileName)
  get(ls()[ls() != "fileName"])
}
gpa <- loadRData("/Users/lukegeel/Downloads/gpa_spring2021.RData")
n <- nrow(gpa)
y <- gpa$Y
x <- gpa$X
linmod <- lm(y~x)
b1 <- linmod$coef[2]
s.b1 <- summary(linmod)$coef[2, 2]
alpha <- 0.01
tquantile <- qt(1 - alpha/2, n - 2)
statistic <- b1/s.b1
tquantile
```

```
## [1] 2.618137
```

```
statistic
```

```
##          x
## 2.716548
```

- c. What is the P-value of your test in part (b)? How does it support the conclusion reached in part (b)? P-value: 0.007 The P-value is less than the alpha level so it supports our conclusion that a linear association exists.

```
library(tidyverse)
loadRData <- function(fileName) {
  load(fileName)
  get(ls()[ls() != "fileName"])
}
gpa <- loadRData("/Users/lukegeel/Downloads/gpa_spring2021.RData")
n <- nrow(gpa)
y <- gpa$Y
x <- gpa$X
linmod <- lm(y~x)
b1 <- linmod$coef[2]
s.b1 <- summary(linmod)$coef[2, 2]
alpha <- 0.01
tquantile <- qt(1 - alpha/2, n - 2)
pvalue <- 2*pt(-abs(b1/s.b1), n - 2)
pvalue
```

```
##          x
## 0.007589142
```

2.5. Refer to Copier maintenance Problem 1.20. a. Estimate the change in the mean service time when the number of copiers serviced increases by one. Use a 90 percent confidence interval. Interpret your confidence interval. We can be 90% confident that when the number of copiers serviced increases by one, the service time increases by between 14.0258 and 16.08582 minutes.

```
data <- loadRData("/Users/lukegeel/Downloads/copier_spring2021.RData")
n <- nrow(data)
y <- data$Y
x <- data$X
linmod <- lm(y~x)
b1 <- linmod$coef[2]
s.b1 <- summary(linmod)$coef[2, 2]
alpha <- 0.1
qt <- qt(alpha/2, n - 2)
pvalue <- pt(-abs(b1/s.b1), n-2) + (1-pt(abs(b1/s.b1), n-2))
lower <- b1+s.b1*qt(alpha/2, n-2)
upper <- b1-s.b1*qt(alpha/2, n-2)
lower
```

```
##          x
## 14.0258
```

```
upper
```

```
##          x
## 16.08582
```

- b. Conduct a t test to determine whether or not there is a linear association between X and Y here; control the alpha risk at .10. State the alternatives, decision rule, and conclusion. What is the P-value of your test? Alternative hypothesis: There is not a linear association between X and Y. Decision rule: If the test statistic exceeds the t-value given the alpha level, we reject the alternative hypothesis. Conclusion: The test statistic doesn't exceed t, so we fail to reject the alternative hypothesis and say that there isn't a linear association between X and Y. P-value: 6.093652e-27

```
data <- loadRData("/Users/lukegeel/Downloads/copier_spring2021.RData")
n <- nrow(data)
y <- data$Y
x <- data$X
linmod <- lm(y~x)
b1 <- linmod$coef[2]
s.b1 <- summary(linmod)$coef[2, 2]
alpha <- 0.1
tquantile <- qt(1 - alpha/2, n - 2)
statistic <- b1/s.b1
pvalue <- 2*pt(-abs(b1/s.b1), n - 2)
pvalue
```

```
##          x
## 6.093652e-27
```

```
tquantile
```

```
## [1] 1.681071
```

```
statistic
```

```
##          x
## 24.57245
```

- c. Are your results in parts (a) and (b) consistent? Explain. No, the results aren't consistent. From a, we get there the mean change is about 15 minutes for each additional copier but from b we concluded that there isn't a linear relationship. These results contradict each other.
- d. Does b0 give any relevant information here about the "start-up" time on calls-Le., about the time required before service work is begun on the copiers at a customer location? No, b0 does not give any relevant information about the start-up time on calls.

1.44. Refer to the CDI data set in Appendix C.2. a. For each geographic region, regress per capita income in a CDr (Y) against the percentage of individuals in a county having at least a bachelor's degree (X). Assume that first-order regression model (1.1) is appropriate for each region. State the estimated regression functions. Region 1:  $9223.8 + 522.2X$  Region 2:  $13581.4 + 238.7X$  Region 3:  $10529.8 + 330.6X$  Region 4:  $8615.1 + 440.3X$

```
cdi <- loadRData("/Users/lukegeel/Downloads/cdi_spring2021.RData")
region1 <- cdi%>%filter(X17 == 1)
n <- nrow(region1)
Y <- region1$X15
X <- region1$X12
linmod1 <- lm(Y~X)
linmod1
```

```
##
## Call:
## lm(formula = Y ~ X)
##
## Coefficients:
## (Intercept)          X
##      9223.8       522.2
```

```
region2 <- cdi%>%filter(X17 == 2)
n <- nrow(region2)
Y <- region2$X15
X <- region2$X12
linmod2 <- lm(Y~X)
linmod2
```

```
##
## Call:
## lm(formula = Y ~ X)
##
## Coefficients:
## (Intercept)          X
##      13581.4       238.7
```

```
region3 <- cdi%>%filter(X17 == 3)
n <- nrow(region3)
Y <- region3$X15
X <- region3$X12
linmod3 <- lm(Y~X)
linmod3
```

```
##
## Call:
## lm(formula = Y ~ X)
##
## Coefficients:
## (Intercept)          X
##      10529.8       330.6
```

```
region4 <- cdi%>%filter(X17 == 4)
n <- nrow(region4)
Y <- region4$X15
X <- region4$X12
linmod4 <- lm(Y~X)
linmod4
```

```
##
## Call:
## lm(formula = Y ~ X)
##
## Coefficients:
## (Intercept)          X
##       8615.1       440.3
```

- b. Are the estimated regression functions similar for the four regions? Discuss. Yes, the estimated regression functions are relatively similar for all four regions. The intercepts are all between 8000 and 14000 and the slope of X are all between 230 and 520.
- c. Calculate MSE for each region. Is the variability around the fitted regression line approximately the same for the four regions? Discuss. Region 1 MSE: 7335008 Region 2 MSE: 4411341 Region 3 MSE: 7474349 Region 4 MSE: 8214318 I believe that the variability around the fitted regression line is approximately the same for the four regions as all the MSEs are very large, all between 4 and 8 million.

```
loadRData("/Users/lukegeel/Downloads/cdi_spring2021.RData")
```

```
##      X1      X2 X3      X4      X5      X6      X7      X8      X9      X10
## 1      1      Los_Angeles CA  4060 8863164 32.1  9.7 23734 27700 688936
## 2      2      Cook IL    946 5105067 29.2 12.4 14246 21550 436936
## 3      3      Harris TX  1729 2818199 31.3  7.1  9137 12449 253526
## 4      4      San_Diego CA  4205 2498016 33.5 10.9  5915  6179 173821
## 5      5      Orange CA    790 2410556 32.6  9.2  6874  6369 144524
## 6      6      Kings NY     71 2300664 28.3 12.4  6097  8942 680966
## 7      7      Maricopa AZ  9204 2122101 29.2 12.5  4200  6104 177593
## 8      8      Wayne MI    614 2111687 27.4 12.5  4819  9490 193978
## 9      9      Dade FL   1945 1937094 27.1 13.9  4312  8840 244725
## 10     10      Dallas TX    880 1852810 32.6  8.2  4665  6934 214258
## 11     11      Philadelphia PA   135 1585577 29.1 15.2  6043 10494 109148
## 12     12      King WA   2126 1507319 30.1 11.1  4050  4009 124959
```

##	13	13	Santa_Clara	CA	1291	1497577	32.6	8.7	3970	3342	77009
##	14	14	San_Bernardino	CA	20062	1418380	30.1	8.8	2482	3349	83110
##	15	15	Cuyahoga	OH	458	1412140	26.3	15.6	4901	8132	73150
##	16	16	Middlesex	MA	824	1398468	31.7	12.5	5645	4152	35825
##	17	17	Allegheny	PA	730	1336449	26.2	17.4	7242	8436	50186
##	18	18	Suffolk	NY	911	1321864	27.9	10.8	2802	3904	66723
##	19	19	Nassau	NY	287	1287348	25.7	14.2	5492	5200	43203
##	20	20	Alameda	CA	738	1279182	30.8	10.6	3839	3284	107338
##	21	21	Broward	FL	1209	1255488	25.3	20.7	6216	5543	107386
##	22	22	Bexar	TX	1247	1185394	29.5	9.9	3215	4086	133098
##	23	23	Riverside	CA	7208	1170413	27.9	13.2	1945	2435	95494
##	24	24	Tarrant	TX	864	1170103	32.2	8.3	2993	3672	132495
##	25	25	Oakland	MI	873	1083592	27.6	10.9	2558	3254	50964
##	26	26	Sacramento	CA	966	1041219	29.7	10.6	2544	2855	84305
##	27	27	Hennepin	MN	557	1032431	31.6	11.3	6381	5395	71753
##	28	28	St._Louis	MO	508	993529	26.1	13.1	1110	1056	42595
##	29	29	Erie	NY	1045	968532	27.3	15.2	2968	4632	55306
##	30	30	Franklin	OH	540	961437	33.5	9.6	2980	4011	82680
##	31	31	Milwaukee	WI	242	959275	29.3	13.6	3360	4141	73681
##	32	32	Westchester	NY	433	874866	26.3	14.4	3985	3540	37118
##	33	33	Hamilton	OH	407	866228	28.0	13.3	3503	4683	57208
##	34	34	Palm_Beach	FL	1974	863518	23.3	24.4	2045	3164	76142
##	35	35	Hartford	CT	736	851783	28.3	14.1	2822	2940	51926
##	36	36	Pinellas	FL	280	851659	22.4	26.0	1913	4458	62344
##	37	37	Honolulu	HI	600	836231	30.6	11.0	1763	2174	51032
##	38	38	Hillsborough	FL	1051	834054	29.4	12.2	1994	3068	89895
##	39	39	Fairfield	CT	626	827645	26.7	13.3	2776	2494	44374
##	40	40	Shelby	TN	755	826330	29.4	10.4	2873	4918	67032
##	41	41	Bergen	NJ	234	825380	25.4	15.3	2641	2279	28521
##	42	42	Fairfax_County	VA	396	818584	29.2	6.5	2960	135	30202
##	43	43	New_Haven	CT	606	804219	28.7	14.7	2970	2486	52903
##	44	44	Contra_Costa	CA	720	803732	26.5	10.9	2699	1781	51243
##	45	45	Marion	IN	396	797159	30.6	11.7	2519	4654	61004
##	46	46	DuPage	IL	334	781666	29.0	8.7	1249	1842	29708
##	47	47	Essex	NJ	126	778206	28.6	12.7	2923	4841	75595
##	48	48	Montgomery	MD	495	757027	28.6	10.2	3735	1507	34754
##	49	49	Clark	NV	7911	741459	29.0	10.5	1047	2011	52786
##	50	50	Baltimore_City	MD	81	736014	30.0	13.7	5759	6203	87355
##	51	51	Prince_George's	MD	486	729268	33.7	6.9	1256	1322	54469
##	52	52	Salt_Lake	UT	737	725956	27.8	8.5	706	2076	58610
##	53	53	San_Francisco	CA	47	723959	32.2	14.5	4854	3640	71234
##	54	54	Macomb	MI	480	717400	28.2	12.3	570	1202	41048
##	55	55	Monroe	NY	659	713968	29.0	12.5	2433	3077	43780
##	56	56	Worcester	MA	1513	709705	29.2	13.7	2338	2205	7099
##	57	57	Baltimore	MD	599	692134	27.8	14.0	1101	641	46789
##	58	58	Montgomery	PA	483	678111	26.1	15.0	2206	2425	20335
##	59	59	Orange	FL	908	677491	33.2	10.6	1894	2929	52577
##	60	60	Duval	FL	774	672971	30.7	10.7	1164	2623	68586
##	61	61	Middlesex	NJ	311	671780	31.6	11.8	1866	1880	30548
##	62	62	Essex	MA	498	670080	27.3	14.1	4066	2009	34312
##	63	63	Ventura	CA	1846	669016	28.8	9.4	1159	1372	30235
##	64	64	Fresno	CA	5963	667490	28.3	10.4	1218	1681	62004
##	65	65	Pima	AZ	9187	666880	28.9	13.7	1480	2016	57051
##	66	66	Jefferson	KY	385	664937	27.0	13.4	2162	3559	32419
##	67	67	Suffolk	MA	59	663906	39.2	12.1	5360	6154	68808
##	68	68	Jefferson	AL	1113	651525	26.7	14.0	2416	4602	55604
##	69	69	San_Mateo	CA	449	649623	28.4	12.3	2284	1642	30473
##	70	70	Fulton	GA	529	648951	31.6	10.0	3481	5757	93025
##	71	71	Jackson	MO	605	633232	28.1	13.1	1688	3762	61760
##	72	72	Norfolk	MA	400	616087	28.8	14.1	1861	1903	14830
##	73	73	District_of_Columbia	DC	61	606900	33.6	12.8	3674	4262	64393
##	74	74	Oklahoma	OK	709	599611	28.4	12.1	1841	3487	57045
##	75	75	Providence	RI	413	596270	29.9	15.8	1474	2360	34627
##	76	76	El_Paso	TX	1013	591610	29.5	8.1	789	1650	54002
##	77	77	Pierce	WA	1676	586203	29.5	10.5	711	1226	41980
##	78	78	Multnomah	OR	435	583887	28.4	13.6	2200	3009	58216
##	79	79	Travis	TX	989	576407	38.0	7.3	1011	1392	60961
##	80	80	Montgomery	OH	462	573809	28.1	12.6	1223	3068	36665
##	81	81	Monmouth	NJ	472	553124	25.6	12.7	1426	1904	22302
##	82	82	Hudson	NJ	47	553099	31.5	12.7	950	2443	40581
##	83	83	Delaware	PA	184	547651	27.6	15.5	1317	1588	18924
##	84	84	De_Kalb	GA	268	545837	32.6	8.5	1106	922	56950
##	85	85	Yamhill	OR	6140	544477	28.2	8.7	644	1104	36210

##	85	85	Kern	CA	8142	5434	11	28.3	9.1	644	1194	36318
##	86	86	Bucks	PA	608	541174	26.6	10.9	1317	1435	16894	
##	87	87	Lake	IL	448	516418	28.9	8.4	1094	1499	22349	
##	88	88	Summit	OH	413	514990	26.8	13.8	1155	2226	26228	
##	89	89	Mecklenburg	NC	527	511433	32.0	9.4	1381	2021	57999	
##	90	90	Davidson	TN	502	510784	32.1	11.6	2398	3847	45237	
##	91	91	Bristol	MA	556	506325	27.6	14.4	1369	1306	22023	
##	92	92	Tulsa	OK	570	503341	28.2	11.5	1655	2512	39496	
##	93	93	Camden	NJ	222	502824	27.5	12.2	1309	2041	34814	
##	94	94	Kent	MI	856	500631	29.6	10.8	1099	1460	31553	
##	95	95	Orleans	LA	181	496938	28.3	13.0	2398	4018	54238	
##	96	96	Union	NJ	103	493819	26.8	15.0	1332	2541	30299	
##	97	97	Ramsey	MN	156	485765	31.0	12.2	745	1140	30574	
##	98	98	San_Joaquin	CA	1399	480628	27.7	11.1	623	1051	41179	
##	99	99	Bernalillo	NM	1166	480577	29.2	10.5	1545	1726	41280	
##	100	100	Lake	IN	497	475594	25.5	12.3	766	2413	29926	
##	101	101	Onondaga	NY	780	468973	29.4	13.0	830	1668	23249	
##	102	102	Denver	CO	153	467610	30.3	13.8	3131	3652	37466	
##	103	103	Snohomish	WA	2090	465642	28.1	9.5	517	672	20323	
##	104	104	Lucas	OH	340	462361	27.8	13.0	1348	3021	38194	
##	105	105	Hampden	MA	619	456310	27.8	14.8	927	1665	24247	
##	106	106	Passaic	NJ	185	453060	28.9	12.9	945	1912	26434	
##	107	107	Jefferson	LA	306	448306	28.1	10.2	525	1648	41625	
##	108	108	Cobb	GA	340	447745	31.7	6.3	1010	983	27582	
##	109	109	New_Castle	DE	426	441946	30.6	11.4	1089	1488	27717	
##	110	110	Jefferson	CO	772	438430	26.9	8.0	485	298	23453	
##	111	111	Plymouth	MA	661	435276	27.1	11.6	575	813	14846	
##	112	112	Ocean	NJ	636	433203	22.4	23.2	591	1475	17379	
##	113	113	Genesee	MI	640	430459	27.2	10.2	686	1830	33136	
##	114	114	Anne_Arundel	MD	416	427239	29.6	8.8	1057	617	21826	
##	115	115	Wake	NC	834	423380	34.5	7.8	1561	1199	26006	
##	116	116	Lancaster	PA	949	422822	27.2	13.1	762	1241	13086	
##	117	117	Morris	NJ	469	421353	26.7	10.6	2123	1599	12147	
##	118	118	Douglas	NE	331	416444	29.0	11.4	757	2889	26006	
##	119	119	Polk	FL	1875	405382	23.9	18.6	520	1288	37290	
##	120	120	Sedgwick	KS	1000	403662	28.5	11.4	1001	1840	34071	
##	121	121	Brevard	FL	1019	398978	26.2	16.6	640	1085	23686	
##	122	122	El_Paso	CO	2127	397014	31.7	8.0	657	1026	25234	
##	123	123	St._Louis_City	MO	62	396685	28.7	16.6	1267	7814	64103	
##	124	124	Burlington	NJ	805	395066	28.8	10.7	697	1150	13034	
##	125	125	VA_Beach_City	VA	248	393069	35.3	5.9	689	530	23412	
##	126	126	Arapahoe	CO	803	391511	28.3	7.4	878	742	27587	
##	127	127	Sonoma	CA	1576	388222	25.8	13.4	812	798	18556	
##	128	128	Hidalgo	TX	1569	383545	26.4	10.1	289	860	26712	
##	129	129	East_Baton_Rouge	LA	456	380105	31.5	9.2	675	1876	41592	
##	130	130	Mobile	AL	1233	378643	26.7	11.8	965	1898	30409	
##	131	131	Chester	PA	756	376396	27.1	10.9	1303	920	9491	
##	132	132	Volusia	FL	1106	370712	24.3	22.8	612	1349	25736	
##	133	133	Stanislaus	CA	1495	370522	27.4	10.8	677	1306	25461	
##	134	134	Westmoreland	PA	1023	370321	23.3	17.1	459	1306	7445	
##	135	135	Santa_Barbara	CA	2739	369608	32.8	12.3	1123	1031	18313	
##	136	136	Stark	OH	576	367585	24.9	14.4	548	1537	17466	
##	137	137	Dane	WI	1202	367085	35.6	9.3	1084	1382	20344	
##	138	138	Spokane	WA	1764	361364	27.0	13.2	838	1346	20042	
##	139	139	Will	IL	837	357313	27.4	8.6	353	746	16432	
##	140	140	Monterey	CA	3322	355660	32.6	9.8	515	602	17870	
##	141	141	Johnson	KS	477	355054	27.5	9.4	1231	925	15238	
##	142	142	Gwinnett	GA	433	352910	32.6	4.7	347	439	17119	
##	143	143	Pulaski	AR	771	349660	28.5	11.5	1463	2785	42404	
##	144	144	Guilford	NC	650	347420	30.4	11.9	918	1188	28212	
##	145	145	Solano	CA	828	340421	29.7	8.2	290	503	21756	
##	146	146	York	PA	905	339574	26.6	13.1	460	951	11292	
##	147	147	Berks	PA	859	336523	26.1	15.6	476	1041	12827	
##	148	148	Hillsborough	NH	877	336073	30.0	10.3	755	1050	12843	
##	149	149	Knox	TN	509	335749	30.0	12.7	1022	2178	22422	
##	150	150	Lee	FL	804	335113	21.5	24.7	509	1202	18442	
##	151	151	Luzerne	PA	891	328149	24.1	19.7	462	1495	4982	
##	152	152	Mercer	NJ	226	325824	29.2	13.0	1069	1724	20153	
##	153	153	Greenville	SC	792	320167	28.2	11.9	785	1358	20504	
##	154	154	Kane	IL	521	317471	27.8	9.3	478	1263	16721	
##	155	155	Tulare	CA	4824	311921	26.3	10.8	401	656	19489	
##	156	156	Washington	OR	724	311554	27.6	10.1	328	294	12630	
##	157	157	Orange	NY	816	307647	28.0	10.4	549	986	10975	

## 158 158	Waukesha	WI	556	304715	24.4	9.8	615	677	8935
## 159 159	Allen	IN	657	300836	27.4	11.4	522	1268	19842
## 160 160	Charleston	SC	917	295039	34.1	10.1	832	1956	28190
## 161 161	Albany	NY	524	292594	30.4	14.7	820	1246	15077
## 162 162	Butler	OH	467	291479	29.9	10.2	334	878	13850
## 163 163	Nueces	TX	836	291145	27.3	10.1	574	1406	28606
## 164 164	Lehigh	PA	347	291130	26.3	15.4	656	1305	12254
## 165 165	Seminole	FL	308	287529	27.9	10.3	378	352	17518
## 166 166	Richland	SC	757	285720	34.7	9.5	966	1207	24101
## 167 167	Hamilton	TN	543	285536	26.3	13.5	723	1573	23532
## 168 168	Washtenaw	MI	710	282937	39.5	7.5	4561	1730	19367
## 169 169	Lane	OR	4554	282912	27.4	13.1	497	654	16091
## 170 170	Ingham	MI	559	281912	37.4	8.7	838	1438	17337
## 171 171	Pasco	FL	745	281131	18.4	32.3	244	941	12509
## 172 172	Clackamas	OR	1868	278850	23.1	11.5	379	345	12855
## 173 173	Sarasota	FL	572	277776	18.2	32.1	774	1363	19801
## 174 174	Erie	PA	802	275572	27.5	13.8	587	1417	9936
## 175 175	Dakota	MN	570	275227	30.6	6.4	209	283	10953
## 176 176	Cumberland	NC	653	274566	37.4	6.2	290	586	25247
## 177 177	Denton	TX	889	273525	36.9	5.0	252	458	20372
## 178 178	Lorain	OH	493	271126	26.4	11.6	292	941	9864
## 179 179	Forsyth	NC	410	265878	29.2	12.3	903	1609	21554
## 180 180	Rockland	NY	174	265475	25.5	10.1	1179	745	7194
## 181 181	Adams	CO	1192	265038	29.6	7.6	449	318	19369
## 182 182	Mahoning	OH	415	264806	23.5	17.1	619	1473	13181
## 183 183	Collin	TX	848	264036	29.8	5.3	389	571	17625
## 184 184	Utah	UT	1998	263590	33.9	7.0	279	544	10605
## 185 185	St._Clair	IL	664	262852	26.8	12.7	434	1088	14563
## 186 186	Escambia	FL	664	262798	29.2	11.9	951	1584	14380
## 187 187	Norfolk_City	VA	54	261229	41.7	10.5	979	1471	25194
## 188 188	Cameron	TX	906	260120	25.9	10.6	265	825	18842
## 189 189	Dutchess	NY	802	259462	29.0	11.4	536	741	9087
## 190 190	New_London	CT	666	254957	31.2	11.9	364	515	7807
## 191 191	Washoe	NV	6343	254667	29.5	10.3	619	990	18831
## 192 192	Hinds	MS	869	254441	29.5	11.2	1253	2118	28841
## 193 193	Winnebago	IL	514	252913	26.5	12.7	602	910	19674
## 194 194	Oneida	NY	1213	250836	27.7	15.5	429	905	9234
## 195 195	Madison	IL	725	249238	26.2	13.9	441	1120	10666
## 196 196	Caddo	LA	882	248253	25.2	13.3	701	1868	22091
## 197 197	Northampton	PA	374	247105	26.9	15.0	434	933	6452
## 198 198	St._Joseph	IN	457	247052	28.2	14.1	416	927	10637
## 199 199	Rockingham	NH	695	245845	29.0	9.2	343	514	7295
## 200 200	Cumberland	ME	836	243135	29.1	13.0	742	1104	13816
## 201 201	Somerset	NJ	305	240279	28.6	10.8	600	374	8308
## 202 202	Jefferson	TX	904	239397	25.8	14.0	446	1724	21677
## 203 203	Madison	AL	805	238912	31.4	8.9	697	933	6635
## 204 204	Clark	WA	628	238053	25.3	10.6	225	299	10706
## 205 205	Dauphin	PA	525	237813	26.8	14.3	816	1425	11563
## 206 206	Marin	CA	520	230096	24.7	12.3	942	488	9460
## 207 207	Gloucester	NJ	325	230082	28.0	10.8	194	339	9746
## 208 208	Santa_Cruz	CA	446	229734	29.9	11.3	427	390	13707
## 209 209	Marion	OR	1185	228483	25.7	14.3	368	498	14825
## 210 210	Trumbull	OH	616	227813	23.9	14.4	293	925	7315
## 211 211	Spartanburg	SC	811	226800	27.0	12.6	275	832	17198
## 212 212	Fort_Bend	TX	875	225421	26.8	4.9	222	301	9433
## 213 213	Fayette	KY	285	225366	34.9	9.9	1187	1851	17378
## 214 214	Boulder	CO	743	225339	34.2	7.6	431	387	14124
## 215 215	Atlantic	NJ	561	224327	29.0	14.5	401	990	25167
## 216 216	Kalamazoo	MI	562	223411	32.4	10.6	422	793	15306
## 217 217	Lubbock	TX	900	222636	34.1	9.8	667	1562	14509
## 218 218	Niagara	NY	523	220756	25.8	15.1	214	893	9437
## 219 219	Lackawanna	PA	459	219039	24.3	19.8	433	1136	4368
## 220 220	Galveston	TX	399	217399	26.9	10.5	249	1592	18586
## 221 221	San_Luis_Obispo	CA	3305	217162	31.4	14.1	410	522	8103
## 222 222	Chatham	GA	440	216935	28.5	12.8	618	1112	18732
## 223 223	Prince_William_County	VA	338	215686	32.3	3.0	189	153	9001
## 224 224	Lake	OH	228	215499	26.4	12.0	180	359	5481
## 225 225	Lancaster	NE	839	213641	33.9	10.8	387	778	16414
## 226 226	St._Charles	MO	561	212907	29.0	6.9	235	613	7785
## 227 227	Broome	NY	707	212160	28.6	15.0	419	816	7435
## 228 228	Saginaw	MI	809	211946	25.5	12.0	496	1193	16190
## 229 229	Manatee	FL	741	211707	21.0	28.1	342	855	16916
## 230 230	Montgomery	AL	790	209085	28.4	11.6	823	1102	17388

##	231	231	Greene	MO	675	207949	31.0	13.3	782	1785	13551
##	232	232	Kanawha	WV	903	207619	23.9	15.7	569	1342	10246
##	233	233	Ada	ID	1055	205775	27.6	10.4	367	557	9701
##	234	234	Washington	PA	857	204584	23.7	17.5	277	687	4526
##	235	235	St._Louis	MN	6226	198213	24.3	16.9	334	1391	7518
##	236	236	Cumberland	PA	550	195257	28.6	13.4	372	733	5247
##	237	237	Marion	FL	1579	194833	21.6	22.1	214	451	14860
##	238	238	Brown	WI	529	194594	29.6	10.8	344	632	8101
##	239	239	Leon	FL	667	192493	38.5	8.2	626	823	23363
##	240	240	Brazoria	TX	1387	191707	28.7	7.8	153	318	8692
##	241	241	Bell	TX	1059	191088	34.6	8.8	251	572	10865
##	242	242	Kitsap	WA	396	189731	27.7	10.7	178	244	8996
##	243	243	Richmond	GA	324	189719	31.1	10.0	336	1787	17918
##	244	244	McLennan	TX	1042	189123	30.0	13.6	267	560	16486
##	245	245	Yakima	WA	4296	188823	25.1	13.0	339	518	15139
##	246	246	Davis	UT	305	187941	26.2	6.1	165	248	6279
##	247	247	Ottawa	MI	566	187768	28.9	9.8	182	313	6140
##	248	248	Howard	MD	252	187328	29.8	6.1	406	208	9057
##	249	249	Barnstable	MA	396	186605	22.1	22.0	338	384	7441
##	250	250	Larimer	CO	2601	186136	32.1	9.6	317	409	8921
##	251	251	Beaver	PA	435	186093	23.6	16.9	240	616	4088
##	252	252	McHenry	IL	604	183241	26.3	9.4	177	371	4854
##	253	253	Peoria	IL	620	182827	26.0	14.2	426	1219	12483
##	254	254	Montgomery	TX	1044	182201	25.5	8.6	166	340	9469
##	255	255	Harford	MD	440	182132	28.8	8.3	289	333	6735
##	256	256	Butte	CA	1640	182120	28.2	17.3	395	625	8939
##	257	257	Clayton	GA	143	182052	32.4	5.8	148	346	15419
##	258	258	Durham	NC	291	181835	33.7	10.7	1253	1496	15477
##	259	259	Alachua	FL	874	181596	40.1	9.3	509	1096	18218
##	260	260	Saratoga	NY	812	181276	28.4	10.3	200	221	5281
##	261	261	Muscogee	GA	216	179278	30.6	10.8	652	1168	11454
##	262	262	Merced	CA	1929	178403	28.2	9.2	187	337	8587
##	263	263	Sangamon	IL	868	178386	25.8	13.8	527	1330	11929
##	264	264	Gaston	NC	357	175093	27.3	12.1	126	368	11865
##	265	265	Racine	WI	333	175034	26.0	12.0	242	532	11110
##	266	266	Buncombe	NC	656	174821	24.8	16.1	513	725	9512
##	267	267	Cleveland	OK	536	174253	33.9	6.7	135	319	12194
##	268	268	Litchfield	CT	920	174092	25.4	14.1	415	411	3593
##	269	269	Champaign	IL	997	173025	41.6	8.8	441	805	11508
##	270	270	Placer	CA	1404	172796	23.5	12.0	265	322	8904
##	271	271	Jefferson	MO	657	171380	28.5	8.3	50	230	3128
##	272	272	Arlington_County	VA	26	170936	37.6	11.3	1289	781	12526
##	273	273	Newport_News_City	VA	68	170045	33.9	9.3	439	836	11776
##	274	274	Calcasieu	LA	1071	168134	26.6	10.9	244	845	6399
##	275	275	Lexington	SC	701	167611	27.8	8.9	205	259	9814
##	276	276	Harrison	MS	581	165365	30.0	10.8	332	764	7043
##	277	277	Ulster	NY	1127	165304	27.8	13.0	257	413	4701
##	278	278	Vanderburgh	IN	235	165058	27.0	15.7	386	1376	8405
##	279	279	Lafayette	LA	270	164762	31.1	8.3	395	1018	10599
##	280	280	York	ME	991	164587	26.4	12.6	242	404	6027
##	281	281	Cambria	PA	688	163029	23.0	18.7	169	892	3187
##	282	282	Wyandotte	KS	151	161993	27.4	13.0	284	1019	18902
##	283	283	Berrien	MI	571	161378	25.2	13.7	352	688	12229
##	284	284	Thurston	WA	727	161238	25.3	11.7	290	500	7882
##	285	285	Kent	RI	170	161135	26.2	15.1	255	359	7302
##	286	286	Shawnee	KS	550	160976	26.2	13.1	328	661	13845
##	287	287	Muskegon	MI	509	158983	25.8	13.0	232	660	12181
##	288	288	Weber	UT	576	158330	26.0	11.1	278	573	9191
##	289	289	Elkhart	IN	464	156198	26.8	11.2	155	478	7573
##	290	290	Rensselaer	NY	654	154429	29.8	13.2	272	616	5297
##	291	291	Clay	MO	397	153411	28.3	10.4	99	693	11085
##	292	292	Schuylkill	PA	779	152585	22.9	20.0	157	634	2119
##	293	293	Lake	FL	953	152104	19.0	27.5	205	664	7099
##	294	294	Collier	FL	2026	152099	22.4	22.8	268	431	9426
##	295	295	Butler	PA	789	152013	27.0	13.5	166	261	3420
##	296	296	Chesapeake_City	VA	341	151976	28.6	8.4	189	210	8427
##	297	297	Smith	TX	929	151309	26.2	13.7	317	795	11712
##	298	298	Tuscaloosa	AL	1325	150522	33.3	11.4	309	731	12377
##	299	299	Frederick	MD	663	150208	28.7	9.4	156	241	4939
##	300	300	Clermont	OH	452	150187	28.1	8.7	87	151	5114
##	301	301	St._Lucie	FL	573	150171	22.9	21.0	161	425	9842
##	302	302	Bibb	GA	250	149967	27.5	12.9	446	1010	12701
##	303	303	Concord	NC	767	148828	30.7	14.4	65	100	2565



##	303	303	Onslow	NC	767	149838	49.7	4.4	65	133	7505
##	304	304	Jackson	MI	707	149756	27.1	12.3	92	573	8630
##	305	305	Schenectady	NY	206	149285	26.1	16.5	325	721	6364
##	306	306	Rock_Island	IL	427	148723	24.9	15.0	231	769	7154
##	307	307	Clark	OH	400	147548	25.3	13.8	176	463	10131
##	308	308	Shasta	CA	3786	147036	22.8	14.1	173	468	7336
##	309	309	Penobscot	ME	3396	146601	30.0	11.5	228	598	4749
##	310	310	Hampshire	MA	529	146568	38.2	11.6	315	236	2547
##	311	311	Jackson	OR	2785	146389	22.1	16.2	262	522	7170
##	312	312	Washington	MN	392	145896	26.5	6.6	121	92	5365
##	313	313	St._Clair	MI	725	145607	25.9	12.3	140	431	6568
##	314	314	Fayette	PA	790	145351	22.9	18.0	103	409	3612
##	315	315	Anderson	SC	718	145196	25.5	13.6	182	456	7525
##	316	316	St._Tammany	LA	854	144508	24.2	8.9	177	512	4447
##	317	317	Horry	SC	1134	144053	28.2	12.7	247	505	12459
##	318	318	Okaloosa	FL	936	143776	30.8	9.3	189	482	5153
##	319	319	Sullivan	TN	413	143596	24.6	14.3	313	982	6236
##	320	320	Middlesex	CT	369	143196	28.4	13.1	296	235	3409
##	321	321	Portage	OH	492	142585	33.6	9.4	116	285	2769
##	322	322	Ouachita	LA	611	142191	28.3	11.2	505	1043	10605
##	323	323	Kenton	KY	163	142031	28.3	11.5	284	733	6925
##	324	324	Chautauqua	NY	1062	141895	25.7	15.7	235	653	5178
##	325	325	Yolo	CA	1012	141092	36.5	9.6	215	168	10650
##	326	326	Outagamie	WI	640	140510	28.2	11.1	203	511	4860
##	327	327	Winnebago	WI	439	140320	30.3	12.8	262	528	6170
##	328	328	Williamson	TX	1124	139551	29.1	7.6	343	185	5724
##	329	329	Rock	WI	721	139510	26.2	12.6	183	491	7643
##	330	330	Berkshire	MA	931	139352	26.0	16.9	329	598	3862
##	331	331	Cumberland	NJ	489	138053	26.7	13.5	107	534	9071
##	332	332	Greene	OH	415	136731	28.5	9.8	151	210	5221
##	333	333	Calhoun	MI	709	135982	25.1	13.3	183	566	9810
##	334	334	Dona_Ana	NM	3807	135510	31.4	8.8	226	240	8850
##	335	335	Hampton_City	VA	52	133793	33.0	9.6	256	251	8376
##	336	336	Monroe	MI	551	133600	26.1	10.4	81	182	6726
##	337	337	Webb	TX	3357	133239	28.5	7.9	107	382	12202
##	338	338	Weld	CO	3993	131821	29.6	10.2	207	281	7901
##	339	339	Chittenden	VT	539	131761	35.8	8.1	696	573	4739
##	340	340	Rapides	LA	1323	131556	26.8	12.0	243	768	6101
##	341	341	York	SC	683	131497	28.7	10.6	152	276	9525
##	342	342	Sussex	NJ	521	130943	26.3	8.9	166	261	3174
##	343	343	Madison	IN	452	130669	25.5	14.0	133	655	5373
##	344	344	Tippecanoe	IN	500	130598	42.3	9.5	421	635	6141
##	345	345	Blair	PA	526	130542	23.7	17.0	139	654	3196
##	346	346	McLean	IL	1184	129180	37.0	10.5	235	588	5949
##	347	347	Porter	IN	418	128932	26.4	9.8	161	379	4014
##	348	348	Tolland	CT	410	128699	33.9	9.0	229	173	1799
##	349	349	Licking	OH	687	128300	26.2	11.8	84	192	1380
##	350	350	Kenosha	WI	273	128181	27.5	12.6	141	334	6616
##	351	351	Whatcom	WA	2120	127780	28.8	12.5	149	214	7070
##	352	352	Bay	FL	764	126994	27.7	12.0	162	478	8634
##	353	353	Davidson	NC	552	126677	26.9	12.0	92	221	5662
##	354	354	Richland	OH	497	126137	25.0	12.9	100	463	7977
##	355	355	El_Dorado	CA	1712	125995	23.0	11.8	150	163	5152
##	356	356	Minnehaha	SD	809	123809	29.7	11.6	376	912	5625
##	357	357	Centre	PA	1108	123786	45.0	9.0	180	270	4136
##	358	358	Tazewell	IL	649	123692	23.9	13.2	95	297	3140
##	359	359	Carroll	MD	449	123372	26.6	10.2	181	123	3430
##	360	360	Pueblo	CO	2389	123051	24.1	15.2	232	555	8640
##	361	361	Wichita	TX	628	122378	29.5	12.8	115	457	10727
##	362	362	Medina	OH	422	122354	24.5	9.7	72	226	563
##	363	363	Brazos	TX	586	121862	49.4	6.7	218	279	8203
##	364	364	Oswego	NY	953	121771	29.8	10.7	89	269	3582
##	365	365	Franklin	PA	772	121082	25.5	14.4	159	296	3155
##	366	366	Mercer	PA	672	121003	24.5	17.2	146	653	2777
##	367	367	Aiken	SC	1073	120940	26.7	11.4	135	191	6835
##	368	368	Hawaii	HI	4028	120317	22.5	12.6	266	391	7226
##	369	369	New_Hanover	NC	199	120284	29.0	12.5	275	554	11892
##	370	370	Merrimack	NH	935	120005	27.7	12.1	201	368	3325
##	371	371	Delaware	IN	393	119659	32.9	12.7	161	494	1064
##	372	372	Taylor	TX	916	119655	30.7	12.0	185	467	6785
##	373	373	Humboldt	CA	3573	119118	27.5	12.3	204	311	5737
##	374	374	Ector	TX	901	118934	27.1	9.3	102	389	14643
##	375	375	Stearns	MN	1345	118791	33.6	10.5	239	661	4101

##	376	376		Lycoming	PA	1235	118710	25.4	15.1	211	668	3826
##	377	377		Rutherford	TN	619	118570	33.1	8.4	120	215	6072
##	378	378		Catawba	NC	400	118412	27.2	11.9	155	464	6830
##	379	379		Macon	IL	581	117206	24.1	14.5	172	725	6103
##	380	380		Pinal	AZ	5370	116379	24.4	13.7	144	309	6275
##	381	381		Calhoun	AL	609	116034	28.8	12.4	126	486	4901
##	382	382		Kennebec	ME	868	115904	26.1	13.4	236	497	4184
##	383	383		Livingston	MI	568	115645	25.3	8.2	82	93	3760
##	384	384		Marathon	WI	1545	115400	25.9	12.7	156	254	3655
##	385	385		Jackson	MS	727	115243	25.9	9.4	110	346	4777
##	386	386		Florence	SC	799	114344	26.2	11.2	202	731	8421
##	387	387		Lebanon	PA	362	113744	25.3	15.0	98	196	2919
##	388	388		Yellowstone	MT	2635	113419	25.6	12.4	262	554	3879
##	389	389		Washington	AR	950	113409	32.0	11.2	310	651	6122
##	390	390		Wood	OH	617	113269	34.5	10.2	78	124	3759
##	391	391		Benton	WA	1703	112560	25.1	10.1	186	278	6249
##	392	392		Boone	MO	685	112379	40.9	8.4	213	1023	5456
##	393	393		St._Lawrence	NY	2686	111974	31.2	12.1	177	378	3851
##	394	394		Bay	MI	444	111723	25.4	13.4	132	415	4849
##	395	395		Comanche	OK	1069	111486	34.5	8.7	134	347	5979
##	396	396		Alexandria_City	VA	15	111183	38.3	10.3	830	662	8537
##	397	397		Kent	DE	591	110993	29.7	10.3	123	193	5846
##	398	398		Charlotte	FL	694	110975	16.6	33.8	140	632	3741
##	399	399		Jefferson	NY	1272	110943	32.7	10.9	116	336	3064
##	400	400		Napa	CA	754	110765	24.5	16.5	554	1019	5056
##	401	401		Rowan	NC	511	110605	26.0	15.2	121	244	5233
##	402	402		Washington	RI	333	110006	31.0	12.3	103	241	3838
##	403	403		Allen	OH	405	109755	26.2	13.4	168	560	4734
##	404	404		Imperial	CA	4175	109303	25.5	10.2	85	221	8042
##	405	405		Monroe	IN	394	108978	45.8	8.6	251	285	1657
##	406	406		Hamilton	IN	398	108936	25.1	8.2	203	122	1699
##	407	407		Columbiana	OH	533	108276	23.4	14.9	86	485	898
##	408	408		Alamance	NC	431	108213	27.3	14.8	94	340	4152
##	409	409		Pitt	NC	652	107924	35.4	9.9	157	583	4603
##	410	410		Hunterdon	NJ	430	107776	25.6	9.5	216	182	2068
##	411	411		Osceola	FL	1322	107728	27.1	13.9	104	291	9665
##	412	412		Yavapai	AZ	8124	107714	18.3	23.8	114	159	3952
##	413	413		La_Porte	IN	598	107066	26.0	13.1	128	519	6021
##	414	414		La_Salle	IL	1135	106913	23.8	17.2	105	504	2982
##	415	415		Yuma	AZ	5514	106895	27.4	13.8	90	197	5414
##	416	416		Midland	TX	900	106611	26.8	9.0	129	333	7546
##	417	417		Randolph	NC	788	106546	27.1	12.2	74	145	2940
##	418	418		Olmsted	MN	653	106470	29.3	10.0	1830	1437	4310
##	419	419		Vigo	IN	403	106107	30.2	15.1	189	576	3435
##	420	420		Clay	FL	601	105986	26.3	8.5	97	277	4560
##	421	421		Androscoggin	ME	470	105259	27.9	13.4	134	527	4020
##	422	422		Robeson	NC	949	105179	26.7	10.7	79	281	4318
##	423	423		Gregg	TX	274	104948	26.4	13.3	166	420	9181
##	424	424		Wayne	NC	553	104666	29.7	10.2	98	263	4682
##	425	425		Strafford	NH	369	104233	34.8	10.7	137	237	3651
##	426	426		Sheboygan	WI	514	103877	25.4	14.6	193	421	4433
##	427	427		Fairfield	OH	506	103461	25.2	11.3	73	195	625
##	428	428		Sumner	TN	529	103281	25.5	10.2	98	259	3285
##	429	429		Cass	ND	1766	102874	34.4	9.8	343	643	3401
##	430	430		Sumter	SC	666	102637	31.6	9.4	92	214	7138
##	431	431		Sarpy	NE	241	102583	30.4	4.8	37	160	2689
##	432	432		Windham	CT	513	102525	28.5	12.5	127	254	1397
##	433	433		Kings	CA	1390	101469	33.7	7.7	82	180	4449
##	434	434		Wayne	OH	555	101461	26.3	11.6	63	155	2377
##	435	435		Charles	MD	461	101154	29.9	6.5	102	104	5279
##	436	436		Hernando	FL	478	101115	16.4	30.7	82	290	4414
##	437	437		Martin	FL	556	100900	20.4	27.5	166	277	5081
##	438	438		Montgomery	TN	539	100498	35.7	7.9	75	188	6537
##	439	439		Maui	HI	1159	100374	26.2	11.3	134	182	7130
##	440	440		Morgan	AL	582	100043	26.3	11.7	124	464	4693

##		X11	X12	X13	X14	X15	X16	X17
##	1	70.0	22.3	11.6	8.0	20786	184230	4
##	2	73.4	22.8	11.1	7.2	21729	110928	2
##	3	74.9	25.4	12.5	5.7	19517	55003	3
##	4	81.9	25.3	8.1	6.1	19588	48931	4
##	5	81.2	27.8	5.2	4.8	24400	58818	4
##	6	63.7	16.6	19.5	9.5	16803	38658	1
##	7	81.5	22.1	8.8	4.9	18042	38287	4

##	8	70.0	13.7	16.9	10.0	17461	36872	2
##	9	65.0	18.8	14.2	8.7	17823	34525	3
##	10	77.1	26.3	10.4	6.1	21001	38911	3
##	11	64.3	15.2	16.1	8.0	16721	26512	1
##	12	88.2	32.8	5.0	4.6	23779	35843	4
##	13	82.0	32.6	5.0	5.5	25193	37728	4
##	14	75.4	14.9	10.3	8.0	16399	23260	4
##	15	74.0	20.1	11.0	5.5	21086	29776	2
##	16	84.3	35.4	4.2	7.3	25312	35398	1
##	17	79.0	22.6	8.7	5.3	20681	27639	1
##	18	82.2	23.0	3.3	7.0	24262	32071	1
##	19	84.2	30.0	2.5	5.1	31679	40782	1
##	20	81.4	28.8	8.1	5.3	22148	28331	4
##	21	76.8	18.8	7.1	7.4	22355	28066	3
##	22	72.7	19.7	16.2	6.7	15508	18383	3
##	23	74.1	14.6	8.4	10.7	17185	20114	4
##	24	79.9	24.0	8.2	6.6	18825	22027	3
##	25	84.6	30.2	4.4	7.3	26884	29131	2
##	26	82.2	23.0	9.8	6.3	18934	19714	4
##	27	88.2	31.6	6.4	4.3	23705	24474	2
##	28	82.3	29.2	4.0	5.1	24219	24062	2
##	29	76.4	20.0	9.4	6.8	18305	17729	1
##	30	81.0	26.6	9.1	4.2	19040	18306	2
##	31	76.3	19.3	12.6	4.9	18431	17680	2
##	32	81.0	35.3	4.7	5.4	33330	29159	1
##	33	75.6	23.7	10.3	4.5	20580	17827	2
##	34	78.8	22.1	6.2	8.4	26798	23141	3
##	35	77.7	25.8	6.0	6.9	24875	21188	1
##	36	78.1	18.5	6.2	6.2	21610	18404	3
##	37	81.2	24.6	5.4	2.3	21307	17818	4
##	38	75.6	20.2	9.5	6.0	16876	14075	3
##	39	81.0	34.2	4.5	5.9	32342	26768	1
##	40	75.1	20.8	14.7	5.4	18430	15229	3
##	41	81.6	31.7	2.7	5.2	32230	26602	1
##	42	91.4	49.0	2.2	3.2	28999	23738	3
##	43	77.5	24.2	6.0	7.3	22197	17851	1
##	44	86.5	31.6	5.5	5.6	25523	20514	4
##	45	76.8	21.4	9.3	5.0	19148	15264	2
##	46	88.6	36.0	1.7	4.8	26772	20927	2
##	47	70.1	24.0	11.3	7.9	24523	19084	1
##	48	90.6	49.9	2.7	3.3	30081	22772	3
##	49	77.3	13.8	7.5	5.8	18625	13810	4
##	50	60.7	15.5	17.8	9.4	17263	12706	3
##	51	83.2	25.5	3.7	5.0	19568	14270	3
##	52	85.3	23.8	7.7	4.5	15399	11179	4
##	53	78.0	35.0	9.7	5.6	28532	20656	4
##	54	76.9	13.5	4.0	9.4	20924	15011	2
##	55	80.1	26.3	7.7	4.4	21641	15451	1
##	56	77.4	22.2	6.3	10.2	19895	14120	1
##	57	78.4	25.0	3.8	5.7	23470	16244	3
##	58	83.8	32.1	2.2	5.0	28462	19300	1
##	59	78.8	21.2	7.8	6.6	17879	12113	3
##	60	76.9	18.4	9.8	6.6	17662	11886	3
##	61	79.4	26.5	3.4	5.7	24896	16725	1
##	62	80.2	25.9	7.5	9.0	22834	15301	1
##	63	79.4	23.0	5.0	7.0	21420	14330	4
##	64	66.2	16.9	16.8	12.6	16365	10923	4
##	65	80.5	23.3	12.0	3.9	15191	10131	4
##	66	74.1	19.3	10.9	6.3	19140	12727	3
##	67	75.4	27.7	14.4	8.7	23150	15369	1
##	68	73.8	19.9	12.7	5.3	18624	12134	3
##	69	84.1	31.3	4.3	4.2	28819	18721	4
##	70	77.8	31.6	15.4	5.3	22819	14808	3
##	71	79.5	20.0	9.8	6.5	18611	11785	2
##	72	88.0	34.4	3.1	7.5	26909	16578	1
##	73	73.1	33.3	13.3	7.7	23603	14325	3
##	74	79.1	22.6	11.3	6.0	17741	10638	3
##	75	67.0	18.3	8.9	9.0	17866	10653	1
##	76	63.7	15.2	22.4	10.8	11545	6830	3
##	77	83.2	17.5	8.7	6.4	16194	9493	4
##	78	82.9	23.7	8.9	5.1	19215	11219	4
##	79	83.4	34.7	10.2	4.8	18340	10571	3
##	80	77.0	20.0	0.0	5.7	10410	10564	0

##	80	77.8	20.0	9.8	5.7	18410	10564	2
##	81	82.8	28.4	3.4	5.8	27391	15151	1
##	82	64.1	19.7	12.4	9.0	18463	10212	1
##	83	81.4	24.8	5.0	5.3	23658	12956	1
##	84	83.9	32.7	7.2	4.4	21005	11465	3
##	85	67.6	13.3	13.7	11.8	15881	8631	4
##	86	82.9	24.8	2.9	6.7	22548	12202	1
##	87	84.7	32.0	3.7	4.6	27378	14138	2
##	88	78.3	19.7	9.5	6.0	18583	9570	2
##	89	81.6	28.3	7.2	4.8	20942	10710	3
##	90	75.9	24.4	10.0	4.6	19505	9963	3
##	91	65.0	15.9	7.4	12.3	18521	9378	1
##	92	81.7	23.7	10.0	6.1	19295	9712	3
##	93	75.5	21.0	8.0	6.9	19930	10021	1
##	94	80.3	20.7	6.9	7.5	18674	9349	2
##	95	68.1	22.4	27.3	6.1	16578	8238	3
##	96	75.2	25.0	5.3	6.9	26248	12962	1
##	97	85.0	28.8	8.1	4.3	20303	9862	2
##	98	68.6	13.2	12.0	12.0	15453	7427	4
##	99	82.1	26.7	10.9	5.2	17518	8419	4
##	100	73.5	12.8	11.7	7.1	16327	7765	2
##	101	80.7	24.4	7.1	5.6	19401	9099	1
##	102	79.2	29.0	13.1	5.4	22156	10360	4
##	103	85.7	19.3	4.9	5.6	18545	8635	4
##	104	76.2	17.0	12.0	8.8	17815	8237	2
##	105	73.6	17.6	10.9	9.6	19073	8703	1
##	106	68.8	18.7	7.6	8.1	21973	9955	1
##	107	76.0	18.8	11.4	5.6	17101	7666	3
##	108	85.8	33.0	3.8	4.1	21933	9820	3
##	109	80.6	25.2	5.0	6.2	22284	9848	1
##	110	89.8	30.7	4.1	4.1	20997	9206	4
##	111	83.8	22.2	5.3	10.4	21500	9358	1
##	112	74.9	15.3	4.3	7.4	20974	9086	1
##	113	76.8	12.8	14.0	12.6	16829	7244	2
##	114	81.1	24.6	3.0	5.0	22797	9740	3
##	115	85.4	35.3	5.5	3.5	20658	8746	3
##	116	70.5	16.7	5.3	5.3	18878	7982	1
##	117	87.0	36.7	1.8	5.1	31520	13281	1
##	118	84.5	24.9	8.1	2.9	19629	8174	2
##	119	68.0	12.9	9.4	10.1	14835	6014	3
##	120	82.4	22.2	8.3	4.5	19276	7781	2
##	121	82.3	20.4	6.3	7.1	17668	7049	3
##	122	88.3	25.8	8.1	6.1	16807	6673	4
##	123	62.8	15.3	20.6	9.0	18113	7185	2
##	124	81.9	23.6	3.1	5.6	23008	9090	1
##	125	88.0	25.5	4.3	5.4	17697	6956	3
##	126	91.5	35.2	4.4	3.9	22507	8812	4
##	127	84.4	24.5	5.2	5.7	22055	8562	4
##	128	46.6	11.5	36.3	17.6	8899	3413	3
##	129	80.5	27.5	15.1	5.4	17881	6797	3
##	130	70.1	15.5	17.5	7.2	14389	5448	3
##	131	84.9	34.7	3.0	4.6	24732	9309	1
##	132	75.4	14.8	7.9	6.9	15648	5801	3
##	133	68.4	13.0	11.4	14.3	15238	5646	4
##	134	77.7	15.4	8.6	6.9	17069	6321	1
##	135	80.0	26.6	7.4	6.0	21902	8095	4
##	136	76.0	14.3	8.9	6.9	16898	6211	2
##	137	88.9	34.2	4.8	3.1	20087	7374	2
##	138	84.4	20.6	9.8	6.2	16365	5914	4
##	139	80.4	18.0	4.5	7.4	18787	6713	2
##	140	72.9	21.5	8.5	10.9	19465	6923	4
##	141	92.9	40.5	2.5	3.3	26156	9287	2
##	142	86.7	29.6	2.9	4.0	19861	7009	3
##	143	79.0	23.5	10.5	5.8	18225	6373	3
##	144	76.1	24.8	7.3	5.4	20349	7070	3
##	145	82.7	18.7	6.0	7.0	17268	5878	4
##	146	72.8	13.9	4.3	6.2	19502	6622	1
##	147	70.0	15.1	5.2	6.7	19655	6614	1
##	148	82.2	26.4	4.0	7.5	22581	7589	1
##	149	74.6	23.9	10.2	4.6	17382	5836	3
##	150	76.9	16.4	6.1	6.4	18877	6326	3
##	151	72.0	13.1	8.3	8.8	16405	5383	1
##	152	77.1	29.5	5.3	5.5	26026	8480	1

##	153	71.6	21.0	7.8	4.9	17874	5723	3
##	154	77.7	21.4	5.0	7.5	21684	6884	2
##	155	60.2	11.8	18.0	17.1	14710	4588	4
##	156	88.2	29.8	4.8	4.4	19932	6210	4
##	157	77.2	19.5	6.4	6.7	19788	6088	1
##	158	88.0	27.1	2.2	4.1	23004	7010	2
##	159	81.2	19.0	5.6	5.9	19123	5753	2
##	160	75.5	22.4	13.8	4.9	16015	4725	3
##	161	80.9	28.3	6.0	4.5	21003	6145	1
##	162	76.0	18.7	7.6	6.6	16750	4882	2
##	163	68.9	17.0	17.2	7.7	15124	4403	3
##	164	74.6	19.6	4.9	6.6	19785	5760	1
##	165	84.6	26.3	5.3	6.1	17885	5142	3
##	166	79.4	28.0	10.1	4.6	17137	4896	3
##	167	72.5	19.7	10.2	5.9	18242	5209	3
##	168	87.2	41.9	6.4	6.0	22782	6446	2
##	169	83.0	22.2	9.4	6.5	15701	4442	4
##	170	83.9	29.2	11.0	6.9	17458	4922	2
##	171	66.9	9.1	7.9	8.3	13944	3920	3
##	172	85.7	23.6	4.7	4.2	19942	5561	4
##	173	81.3	21.9	4.6	5.1	24948	6930	3
##	174	77.5	16.2	9.5	7.1	16331	4500	1
##	175	90.7	27.6	3.3	4.7	21123	5814	2
##	176	80.3	16.6	12.1	6.4	12923	3548	3
##	177	86.8	32.3	4.5	5.6	17801	4869	3
##	178	75.3	12.3	9.1	9.5	16006	4340	2
##	179	77.6	24.1	7.8	4.4	20645	5489	3
##	180	83.3	33.0	4.1	5.3	26757	7103	1
##	181	78.8	13.0	8.8	5.0	16116	4271	4
##	182	74.6	14.0	12.8	6.7	16256	4305	2
##	183	88.3	39.1	3.9	5.9	22303	5889	3
##	184	87.9	26.2	10.6	4.3	11467	3023	4
##	185	72.6	14.7	13.9	8.4	16190	4256	2
##	186	76.2	18.2	13.3	5.8	15392	4045	3
##	187	72.7	16.8	15.1	6.4	16412	4287	3
##	188	50.0	12.0	33.7	12.5	9728	2530	3
##	189	79.8	24.8	3.6	5.1	22173	5753	1
##	190	80.9	21.8	4.7	6.8	20259	5165	1
##	191	82.5	20.7	6.3	5.0	21327	5431	4
##	192	75.2	26.4	16.7	6.3	16215	4126	3
##	193	76.3	16.7	7.7	7.9	18376	4648	2
##	194	75.1	16.7	8.8	6.7	16477	4133	1
##	195	75.8	14.4	8.5	7.9	17980	4481	2
##	196	73.4	18.2	19.1	7.1	16337	4056	3
##	197	73.1	16.7	4.9	6.8	18336	4531	1
##	198	76.1	19.2	7.1	5.7	17211	4252	2
##	199	86.2	25.9	3.2	8.0	21770	5352	1
##	200	85.0	27.6	5.7	5.7	21362	5194	1
##	201	86.3	38.3	1.4	4.1	33180	7972	1
##	202	74.4	15.5	15.5	6.7	17418	4170	3
##	203	80.2	30.1	8.4	5.2	18990	4537	3
##	204	83.9	16.8	7.0	6.6	16790	3997	4
##	205	77.6	18.6	7.2	5.5	18348	4363	1
##	206	91.9	44.0	3.0	4.0	37541	8638	4
##	207	77.5	18.1	4.7	7.1	18523	4262	1
##	208	81.9	29.7	6.2	8.0	22025	5060	4
##	209	78.7	17.5	9.4	5.7	16022	3661	4
##	210	75.2	11.4	9.4	7.9	16144	3678	2
##	211	63.0	14.3	8.8	5.5	15776	3578	3
##	212	80.9	30.2	6.9	3.8	18301	4125	3
##	213	80.2	30.6	10.2	3.8	19320	4354	3
##	214	91.3	42.1	5.6	3.5	21421	4827	4
##	215	72.9	16.4	6.4	8.3	24035	5392	1
##	216	83.4	27.1	8.9	5.9	18288	4086	2
##	217	74.2	23.4	13.2	5.6	15443	3438	3
##	218	75.8	13.6	8.6	8.3	16647	3675	1
##	219	73.3	14.8	7.8	7.9	16963	3716	1
##	220	75.8	19.3	12.5	7.0	17744	3858	3
##	221	83.3	22.9	6.8	5.8	17221	3740	4
##	222	73.7	18.6	13.6	4.3	17776	3856	3
##	223	87.8	27.6	2.3	4.3	20543	4431	3
##	224	81.1	17.5	3.6	5.7	19692	4244	2
##	225	88.1	27.6	6.0	2.2	17816	3806	2

##	226	83.3	21.2	3.5	6.4	18753	3993	2
##	227	78.9	20.7	6.5	5.9	18058	3831	1
##	228	74.8	13.0	14.7	8.8	16904	3583	2
##	229	75.6	15.5	6.8	5.8	17997	3810	3
##	230	75.3	24.2	14.1	5.9	17469	3653	3
##	231	78.9	20.7	9.2	4.8	16630	3458	2
##	232	72.4	17.6	12.3	7.5	17192	3569	3
##	233	87.2	24.9	6.2	4.1	18786	3866	4
##	234	73.2	13.6	9.7	7.0	16625	3401	1
##	235	80.3	17.3	9.7	6.5	15419	3056	2
##	236	81.0	22.9	2.8	4.6	19254	3759	1
##	237	69.6	11.5	10.8	8.6	13802	2689	3
##	238	82.6	17.7	6.7	4.6	18490	3598	2
##	239	84.9	37.1	9.4	3.9	16422	3161	3
##	240	75.5	15.1	7.7	5.6	17951	3441	3
##	241	79.1	17.2	12.2	6.7	13536	2587	3
##	242	86.6	19.8	7.5	5.2	17009	3227	4
##	243	70.9	17.3	14.8	4.9	15941	3024	3
##	244	71.6	16.6	13.9	6.4	14925	2823	3
##	245	66.1	13.7	15.6	11.6	15374	2903	4
##	246	89.9	23.5	5.5	4.5	13394	2517	4
##	247	79.8	18.7	3.8	6.5	18360	3447	2
##	248	91.1	46.9	2.2	4.1	27546	5160	3
##	249	88.4	28.1	5.8	11.1	23267	4342	1
##	250	88.6	32.3	6.6	4.1	17140	3190	4
##	251	74.9	11.9	10.8	9.7	15162	2822	1
##	252	84.5	21.0	2.5	7.1	21855	4005	2
##	253	77.9	19.5	11.3	7.2	18342	3353	2
##	254	75.5	19.4	9.4	5.5	17084	3113	3
##	255	81.6	21.5	4.1	6.7	20941	3814	3
##	256	77.6	19.5	12.2	9.4	15051	2741	4
##	257	77.2	14.7	7.3	5.8	16171	2944	3
##	258	78.9	33.4	8.7	3.6	19238	3498	3
##	259	82.7	34.6	14.4	4.2	16058	2916	3
##	260	83.0	25.2	4.4	5.9	18857	3418	1
##	261	71.5	16.6	14.9	4.6	15505	2780	3
##	262	63.1	12.0	15.4	14.6	13961	2491	4
##	263	81.8	22.4	7.2	4.8	19601	3497	2
##	264	60.9	10.8	8.2	6.2	16319	2857	3
##	265	76.4	16.5	7.9	6.5	18426	3225	2
##	266	74.5	19.1	8.2	4.9	16934	2960	3
##	267	83.9	25.9	7.0	4.8	14443	2517	3
##	268	80.9	25.0	2.4	7.0	25161	4380	1
##	269	87.5	34.1	8.0	4.5	16957	2934	2
##	270	85.1	22.7	5.3	6.8	20168	3485	4
##	271	71.6	9.0	6.0	8.7	15896	2724	2
##	272	87.5	52.3	4.3	3.6	30242	5169	3
##	273	79.3	18.4	12.2	6.5	15327	2606	3
##	274	70.3	14.7	15.5	7.8	14968	2517	3
##	275	77.3	21.0	6.3	4.1	18126	3038	3
##	276	74.7	16.3	15.4	6.7	13691	2264	3
##	277	76.6	21.6	5.6	5.9	18824	3112	1
##	278	75.2	16.0	9.1	5.7	18093	2986	2
##	279	73.3	22.5	16.2	5.0	16868	2779	3
##	280	79.5	19.0	4.7	6.9	17908	2947	1
##	281	71.2	10.8	11.2	9.2	14473	2360	1
##	282	69.9	10.3	13.9	7.4	14134	2290	2
##	283	74.7	16.7	11.6	9.1	16232	2619	2
##	284	86.5	24.7	7.1	5.9	17312	2791	4
##	285	76.8	20.5	3.7	8.6	20086	3237	1
##	286	84.4	22.3	7.4	4.7	19558	3148	2
##	287	74.2	11.1	12.4	12.0	14767	2348	2
##	288	82.5	18.0	7.8	5.7	15301	2423	4
##	289	72.8	14.2	5.3	6.4	16770	2619	2
##	290	77.7	19.5	6.6	6.3	17774	2745	1
##	291	84.7	20.0	4.2	5.7	18395	2822	2
##	292	68.4	8.1	7.7	9.7	15853	2419	1
##	293	70.6	12.7	7.9	8.9	17496	2661	3
##	294	79.0	22.3	6.4	7.5	25589	3892	3
##	295	78.6	15.6	7.1	6.8	17251	2622	1
##	296	77.1	16.9	7.0	6.1	16924	2572	3
##	297	75.7	19.8	12.6	6.6	17511	2650	3
##	298	68.6	22.0	12.5	5.4	15112	2325	2

##	298	69.6	20.0	13.5	5.4	15113	2215	3
##	299	80.4	22.0	3.5	6.0	19954	2997	3
##	300	72.8	14.5	7.4	6.0	16231	2438	2
##	301	71.7	13.1	8.5	13.8	14137	2123	3
##	302	68.2	17.0	15.9	4.1	17548	2632	3
##	303	83.0	13.4	9.8	5.6	10190	1527	3
##	304	77.7	12.9	9.4	9.9	15750	2359	2
##	305	80.7	23.0	5.7	5.6	20679	3087	1
##	306	77.4	15.0	10.2	7.5	17818	2650	2
##	307	73.4	12.2	10.4	6.5	16676	2461	2
##	308	78.4	13.7	11.0	10.3	16277	2393	4
##	309	79.1	17.7	9.5	8.0	15521	2275	1
##	310	83.0	31.9	5.7	7.7	17853	2617	1
##	311	80.1	17.6	9.7	7.6	15582	2281	4
##	312	90.0	26.2	3.6	4.6	20682	3017	2
##	313	74.8	10.7	9.2	11.8	17480	2545	2
##	314	67.8	9.3	17.6	9.4	14051	2042	1
##	315	64.0	12.9	8.6	6.6	14205	2063	3
##	316	76.9	23.1	11.1	6.2	17129	2475	3
##	317	74.3	16.0	11.6	7.7	14693	2117	3
##	318	83.8	21.0	7.8	6.7	15803	2272	3
##	319	66.8	15.6	10.4	4.4	15747	2261	3
##	320	82.6	28.2	2.6	5.8	24132	3456	1
##	321	79.3	17.6	7.6	6.0	16031	2286	2
##	322	71.6	18.9	19.6	6.4	13869	1972	3
##	323	74.4	17.0	8.2	5.4	16935	2405	3
##	324	74.4	14.2	9.9	7.7	15197	2156	1
##	325	79.1	30.3	9.8	7.2	19727	2783	4
##	326	81.5	16.7	4.6	4.9	17182	2414	2
##	327	80.6	18.2	5.3	4.8	17645	2476	2
##	328	81.4	24.6	7.6	3.8	14934	2084	3
##	329	78.2	13.3	7.8	9.7	16742	2336	2
##	330	77.9	20.9	6.3	10.1	20068	2797	1
##	331	63.4	10.8	10.2	10.7	16819	2322	1
##	332	82.4	26.0	7.1	5.2	18161	2483	2
##	333	76.8	13.8	11.0	8.5	15944	2168	2
##	334	70.4	21.9	20.7	7.4	11379	1542	4
##	335	79.7	19.1	8.8	6.8	14743	1973	3
##	336	74.1	10.5	6.9	11.1	17278	2308	2
##	337	47.8	11.1	33.1	9.8	8973	1196	3
##	338	74.9	18.4	10.6	4.5	15874	2093	4
##	339	86.7	34.0	5.0	4.4	19940	2627	1
##	340	69.0	14.6	18.6	7.1	14615	1923	3
##	341	67.5	16.9	7.6	5.5	16713	2198	3
##	342	85.1	24.9	2.5	6.8	24405	3196	1
##	343	73.5	11.7	10.3	7.5	16018	2093	2
##	344	85.2	30.7	6.8	3.2	15847	2070	2
##	345	75.0	10.5	11.0	9.0	14779	1929	1
##	346	84.7	29.0	5.9	4.6	18961	2449	2
##	347	82.4	18.5	4.6	4.5	17566	2265	2
##	348	84.7	29.2	2.7	5.9	21944	2824	1
##	349	76.4	13.0	8.3	6.3	16412	2106	2
##	350	75.1	12.7	7.6	6.4	17338	2222	2
##	351	83.2	22.0	7.4	6.3	16002	2045	4
##	352	74.7	15.7	11.2	8.0	14814	1881	3
##	353	64.2	10.0	7.3	5.9	15079	1910	3
##	354	73.5	11.6	8.4	9.4	16191	2042	2
##	355	85.9	20.8	5.8	6.1	19250	2425	4
##	356	83.1	21.3	5.4	2.5	18526	2294	2
##	357	83.6	32.3	6.2	6.1	15476	1916	1
##	358	78.6	13.6	7.1	7.4	18008	2227	2
##	359	78.5	19.6	2.5	5.7	22002	2714	3
##	360	73.9	14.0	16.7	7.0	14197	1747	4
##	361	75.1	16.5	12.5	6.8	17119	2095	3
##	362	82.4	18.0	4.1	5.8	18892	2312	2
##	363	79.8	35.8	14.9	3.6	12641	1540	3
##	364	74.7	12.9	8.1	10.0	14834	1806	1
##	365	69.4	12.4	5.7	6.7	16281	1971	1
##	366	75.1	13.6	10.1	7.2	15177	1836	1
##	367	70.7	17.2	11.0	4.9	17898	2165	3
##	368	77.7	18.5	10.9	4.0	16728	2013	4
##	369	78.1	21.2	9.9	5.9	17119	2059	3
##	370	83.2	25.4	3.6	7.1	20600	2472	1

##	371	74.5	16.5	10.3	6.1	15697	1878	2
##	372	75.4	20.7	11.2	5.9	16021	1917	3
##	373	80.5	20.0	12.8	8.8	16138	1922	4
##	374	66.9	11.4	16.6	6.6	14766	1756	3
##	375	78.3	17.5	7.3	5.4	14757	1753	2
##	376	74.5	12.3	8.4	8.0	15778	1873	1
##	377	73.9	18.7	7.6	5.4	15501	1838	3
##	378	66.7	14.2	4.8	6.4	17396	2060	3
##	379	76.2	14.8	9.8	8.9	18021	2112	2
##	380	65.5	8.2	18.7	7.9	11396	1326	4
##	381	67.4	14.2	11.7	7.3	13776	1598	3
##	382	78.9	18.1	7.3	6.8	17131	1986	1
##	383	85.6	19.6	3.0	8.1	21153	2446	2
##	384	75.9	13.5	5.5	5.9	16305	1882	2
##	385	74.4	14.4	14.0	7.1	13475	1553	3
##	386	64.3	14.8	15.9	5.9	14961	1711	3
##	387	70.0	11.8	5.2	5.9	16500	1877	1
##	388	83.7	21.5	9.0	5.3	17272	1959	4
##	389	73.2	20.0	9.8	3.7	14736	1671	3
##	390	83.8	21.9	5.0	6.6	17522	1985	2
##	391	83.9	23.3	8.9	6.8	17332	1951	4
##	392	84.8	36.5	9.3	3.2	17175	1930	2
##	393	73.1	15.1	12.8	9.8	12704	1423	1
##	394	74.0	11.0	10.2	9.2	16499	1843	2
##	395	81.1	18.4	13.3	6.6	13228	1475	3
##	396	86.9	48.5	4.7	4.6	31699	3524	3
##	397	73.1	15.0	8.7	7.6	14946	1659	1
##	398	75.7	13.4	5.2	7.2	16362	1816	3
##	399	76.4	13.6	9.5	11.0	15205	1687	1
##	400	80.7	22.3	4.6	5.9	22668	2511	4
##	401	66.0	11.7	6.8	5.4	15691	1736	3
##	402	82.8	29.1	3.7	6.7	19449	2140	1
##	403	76.1	11.4	10.4	7.3	16542	1816	2
##	404	53.2	9.7	20.8	21.3	14523	1587	4
##	405	82.1	32.9	9.5	3.7	14266	1555	2
##	406	88.7	36.2	2.6	2.8	25681	2798	2
##	407	71.8	8.5	13.0	6.7	12597	1364	2
##	408	67.9	14.6	6.0	4.5	17306	1873	3
##	409	71.0	21.9	15.5	5.5	15852	1711	3
##	410	85.9	34.6	1.8	4.1	30255	3261	1
##	411	73.7	11.2	6.9	6.7	16451	1772	3
##	412	78.9	17.7	9.8	5.4	13681	1474	4
##	413	73.9	11.7	7.8	5.8	16655	1783	2
##	414	73.1	10.5	8.7	9.6	16119	1723	2
##	415	64.9	12.7	15.4	17.8	11490	1228	4
##	416	76.8	26.4	11.5	5.4	19345	2062	3
##	417	62.0	9.1	6.5	4.8	14721	1568	3
##	418	88.0	29.5	4.5	3.3	20515	2184	2
##	419	76.0	18.1	10.7	5.0	15036	1595	2
##	420	81.2	17.9	5.4	5.9	16029	1699	3
##	421	71.8	12.6	8.7	9.7	16154	1700	1
##	422	57.0	11.0	20.7	8.9	10849	1141	3
##	423	75.8	17.7	13.4	8.2	16775	1761	3
##	424	71.2	12.7	11.9	7.3	13350	1397	3
##	425	79.8	21.7	5.1	6.5	17182	1791	1
##	426	77.4	13.8	4.5	5.6	18061	1876	2
##	427	78.8	15.5	6.8	6.3	16342	1691	2
##	428	70.6	14.4	7.0	6.8	16514	1706	3
##	429	87.1	26.5	7.3	2.5	16275	1674	2
##	430	69.8	15.0	16.9	9.4	11803	1211	3
##	431	91.0	25.4	3.5	2.6	16137	1655	2
##	432	71.1	16.8	6.0	9.2	18070	1853	1
##	433	65.6	9.0	15.0	12.8	13907	1411	4
##	434	73.6	13.9	8.4	5.9	16464	1670	2
##	435	81.0	16.2	3.7	4.9	19317	1954	3
##	436	70.5	9.7	7.9	8.2	13919	1407	3
##	437	79.7	20.3	5.0	9.8	27125	2737	3
##	438	77.9	16.5	10.8	8.0	13169	1323	3
##	439	77.0	17.8	5.7	3.2	18504	1857	4
##	440	69.4	15.5	9.4	7.1	16458	1647	3



```
s1 <- summary(linmod1)$s
s1^2
```

```
## [1] 7335008
```

```
s2 <- summary(linmod2)$s
s2^2
```

```
## [1] 4411341
```

```
s3 <- summary(linmod3)$s
s3^2
```

```
## [1] 7474349
```

```
s4 <- summary(linmod4)$s
s4^2
```

```
## [1] 8214318
```

2.63 Refer to the CDI data set in Appendix C.2 and Project I.44. Obtain a separate interval estimate of  $b_1$ , for each region. Use a 90 percent confidence coefficient in each case. Do the regression lines for the different regions appear to have similar slopes? Region 1  $b_1$  CI: (460.5177,583.8) Region 2  $b_1$  CI: (193.4858,283.853) Region 3  $b_1$  CI: (285.7076,375.5158) Region 4  $b_1$  CI: (364.7585,515.8729) Yes, the regression lines for the different regions appear to have similar slopes.

```
n <- nrow(region1)
Y <- region1$X15
X <- region1$X12
linmod1 <- lm(Y~X)
b1 <- linmod1$coef[2]
s.b1 <- summary(linmod1)$coef[2, 2]
alpha <- 0.1
qt <- qt(alpha/2, n - 2)
pvalue <- pt(-abs(b1/s.b1),n-2)+(1-pt(abs(b1/s.b1),n-2))
lower <- b1+s.b1*qt(alpha/2,n-2)
upper <- b1-s.b1*qt(alpha/2,n-2)
lower
```

```
##          X
## 460.5177
```

```
upper
```

```
##          X
## 583.8
```

```
n <- nrow(region2)
Y <- region2$X15
X <- region2$X12
linmod2 <- lm(Y~X)
b1 <- linmod2$coef[2]
s.b1 <- summary(linmod2)$coef[2, 2]
alpha <- 0.1
qt <- qt(alpha/2, n - 2)
pvalue <- pt(-abs(b1/s.b1),n-2)+(1-pt(abs(b1/s.b1),n-2))
lower <- b1+s.b1*qt(alpha/2,n-2)
upper <- b1-s.b1*qt(alpha/2,n-2)
lower
```

```
##          X
## 193.4858
```

```
upper
```

```
##          X
## 283.853
```

```
n <- nrow(region3)
Y <- region3$X15
X <- region3$X12
linmod3 <- lm(Y~X)
b1 <- linmod3$coef[2]
s.b1 <- summary(linmod3)$coef[2, 2]
alpha <- 0.1
qt <- qt(alpha/2, n - 2)
pvalue <- pt(-abs(b1/s.b1), n-2) + (1-pt(abs(b1/s.b1), n-2))
lower <- b1+s.b1*qt(alpha/2, n-2)
upper <- b1-s.b1*qt(alpha/2, n-2)
lower
```

```
##          X
## 285.7076
```

```
upper
```

```
##          X
## 375.5158
```

```
n <- nrow(region4)
Y <- region4$X15
X <- region4$X12
linmod4 <- lm(Y~X)
b1 <- linmod4$coef[2]
s.b1 <- summary(linmod4)$coef[2, 2]
alpha <- 0.1
qt <- qt(alpha/2, n - 2)
pvalue <- pt(-abs(b1/s.b1), n-2) + (1-pt(abs(b1/s.b1), n-2))
lower <- b1+s.b1*qt(alpha/2, n-2)
upper <- b1-s.b1*qt(alpha/2, n-2)
lower
```

```
##          X
## 364.7585
```

```
upper
```

```
##          X
## 515.8729
```

6.29. Refer to the CDI data set in Appendix C.2. a. For each geographic region, regress the number of serious crimes in a CDI (Y) against population density (X" total population divided by land area), per capita personal income (X2 ), and percent high school graduates (X3 ). Use first-order regression model (6.5) with three predictor variables. State the estimated regression functions. Region 1:  $Y = -64466.231 + 17.383X_1 - 1.406X_2 + 1182.577X_3$  Region 2:  $Y = -4163.2673 + 33.6193X_1 + 0.1024X_2 - 2.7616X_3$  Region 3:  $Y = 38862.667 + 5.537 + 1.957X_2 - 670.884X_3$  Region 4:  $Y = 129323.415 + 5.717X_1 + 4.342X_2 - 2159.920X_3$

```
region1 <- cdi%>%filter(X17 == 1)
n <- nrow(region1)
Y <- region1$X10
X1 <- region1$X5/region1$X4
X2 <- region1$X15
X3 <- region1$X11
linmod1 <- lm(Y~X1+X2+X3)
linmod1
```

```
##
## Call:
## lm(formula = Y ~ X1 + X2 + X3)
##
## Coefficients:
## (Intercept)          X1          X2          X3
## -64466.231      17.383      -1.406      1182.577
```

```
region2 <- cdi%>%filter(X17 == 2)
n <- nrow(region2)
Y <- region2$X10
X1 <- region2$X5/region2$X4
X2 <- region2$X15
X3 <- region2$X11
linmod2 <- lm(Y~X1+X2+X3)
linmod2
```

```
##
## Call:
## lm(formula = Y ~ X1 + X2 + X3)
##
## Coefficients:
## (Intercept)          X1          X2          X3
## -4163.2673      33.6193      0.1024     -2.7616
```

```
region3 <- cdi%>%filter(X17 == 3)
n <- nrow(region3)
Y <- region3$X10
X1 <- region3$X5/region3$X4
X2 <- region3$X15
X3 <- region3$X11
linmod3 <- lm(Y~X1+X2+X3)
linmod3
```

```
##
## Call:
## lm(formula = Y ~ X1 + X2 + X3)
##
## Coefficients:
## (Intercept)          X1          X2          X3
##  38862.667      5.537      1.957     -670.884
```

```
region4 <- cdi%>%filter(X17 == 4)
n <- nrow(region4)
Y <- region4$X10
X1 <- region4$X5/region4$X4
X2 <- region4$X15
X3 <- region4$X11
linmod4 <- lm(Y~X1+X2+X3)
linmod4
```

```
##
## Call:
## lm(formula = Y ~ X1 + X2 + X3)
##
## Coefficients:
## (Intercept)          X1          X2          X3
## 129323.415      5.717      4.342    -2159.920
```

- b. Are the estimated regression functions similar for the four regions? Discuss. No, the estimated regression functions are not similar for the four regions. The intercepts are a wide range, some positive some negative. Also, the coefficients aren't always all positive and negative, they are different for each region.
- c. Calculate MSE and R2 for each region. Are these measures similar for the four regions? Discuss. Region 1: MSE: 787397480 R2: 0.8352049 Region 2: MSE: 1087623839 R2: 0.5285125 Region 3: MSE: 1367108167 R2: 0.09250839 Region 4: MSE: 6694591439 R2: 0.08665358 These measurements are not all similar for the four regions. The R2 values range from 0.83 to 0.08 so some are good fits some aren't.

```

region1 <- cdi%>%filter(X17 == 1)
n <- nrow(region1)
Y <- region1$X10
X1 <- region1$X5/region1$X4
X2 <- region1$X15
X3 <- region1$X11
linmod1 <- lm(Y~X1+X2+X3)
#linmod1
region2 <- cdi%>%filter(X17 == 2)
n <- nrow(region2)
Y <- region2$X10
X1 <- region2$X5/region2$X4
X2 <- region2$X15
X3 <- region2$X11
linmod2 <- lm(Y~X1+X2+X3)
#linmod2
region3 <- cdi%>%filter(X17 == 3)
n <- nrow(region3)
Y <- region3$X10
X1 <- region3$X5/region3$X4
X2 <- region3$X15
X3 <- region3$X11
linmod3 <- lm(Y~X1+X2+X3)
#linmod3
region4 <- cdi%>%filter(X17 == 4)
n <- nrow(region4)
Y <- region4$X10
X1 <- region4$X5/region4$X4
X2 <- region4$X15
X3 <- region4$X11
linmod4 <- lm(Y~X1+X2+X3)
#linmod4
s1 <- summary(linmod1)$s
r.sql <- summary(linmod1)$r.sq
s1^2

```

```
## [1] 787397480
```

```
r.sql
```

```
## [1] 0.8352049
```

```

s2 <- summary(linmod2)$s
r.sql2 <- summary(linmod2)$r.sq
s2^2

```

```
## [1] 1087623839
```

```
r.sql2
```

```
## [1] 0.5285125
```

```

s3 <- summary(linmod3)$s
r.sql3 <- summary(linmod3)$r.sq
s3^2

```

```
## [1] 1367108167
```

```
r.sql3
```

```
## [1] 0.09250839
```

```
s4 <- summary(linmod4)$s
r.sq4 <- summary(linmod4)$r.sq
r.sq4
```

```
## [1] 0.08665358
```

```
s4^2
```

```
## [1] 6694591439
```

628. Refer to the CDI data set in Appendix C.2. You have been asked to evaluate two alternative models for predicting the number of active physicians (Y) in a CDI. Proposed model I includes as predictor variables total population (X1), land area (X2), and total personal income (X3). Proposed model II includes as predictor variables population density (X1, total population divided by land area), percent of population greater than 64 years old (X2), and total personal income (X3). (a) Obtain a point estimate and 90% confidence interval for  $\beta_1$  under a normal errors linear regression model that uses the number of active physicians in a CDI (Y) as a response and total population (X1) as a predictor. CI: (0.002762953, 0.002928969)

```
n <- nrow(cdi)
Y <- cdi$X8
X <- cdi$X5
linmod <- lm(Y~X)
b1 <- linmod$coef[2]
s.b1 <- summary(linmod)$coef[2, 2]
alpha <- 0.1
qt <- qt(alpha/2, n - 2)
pvalue <- pt(-abs(b1/s.b1), n-2) + (1-pt(abs(b1/s.b1), n-2))
lower <- b1+s.b1*qt(alpha/2, n-2)
upper <- b1-s.b1*qt(alpha/2, n-2)
lower
```

```
## X
## 0.002762953
```

```
upper
```

```
## X
## 0.002928969
```

b. Obtain a point estimate and 90% confidence interval for  $\beta_1$  under a normal errors linear regression model that uses the number of active physicians in a CDI (Y) as a response and total population (X1), land area (X2), and total personal income (X3) as predictors. CI: (0.0004919585, 0.001485982)

```
n <- nrow(cdi)
Y <- cdi$X8
X1 <- cdi$X5
X2 <- cdi$X4
X3 <- cdi$X16
linmod <- lm(Y~X1+X2+X3)
b1 <- linmod$coef[2]
s.b1 <- summary(linmod)$coef[2, 2]
alpha <- 0.1
qt <- qt(alpha/2, n - 2)
pvalue <- pt(-abs(b1/s.b1), n-2) + (1-pt(abs(b1/s.b1), n-2))
lower <- b1+s.b1*qt(alpha/2, n-2)
upper <- b1-s.b1*qt(alpha/2, n-2)
lower
```

```
## X1
## 0.0004919585
```

```
upper
```

```
## X1
## 0.001485982
```

- c. Compare the interpretations of  $b_1$  in (a) and (c). Do they differ? If so, how? In part a, its predicting the number of physicians based soley on total population where as in part b they number of physicians is predicted by total population, land area, and total personal income.
- d. Compare the point estimates and 90% confidence intervals for  $\beta_1$  obtained in (a) and (b). Are they identical? They are not identical. The CI for part a is much smaller than part b. That means that when predicting the number of physicians using just the total population will produce more accurate results than using total population, land area, and total personal income.
- e. Explain how what you observe in (d) makes sense given what you concluded in (c). Because we used more predictors in part b, the range of the CI would be larger as there are more variables in the prediction.