

# STA 108 Project

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Introduction: The interest of this project is to make inferences about the number of active physicians in terms of the following predictor variables: total population, number of hospital beds and total personal income. Moreover, we also split the dataset based on four different regions and conduct analysis on their relationships. In this report, we use fitting regression models, ANOVA models, confidence intervals, residual plots and Q-Q plots to make analysis on the CDI dataset that provides us information about populous counties in the US.

In this project, we first find the estimated regression functions, then, we plot them on separate graphs and check whether a linear regression relation provides a good fit for each of the three predictor variables. Secondly, we measure and compare the MSE because we want to check which predictor variable has the smallest variability around the fitted regression line. Thirdly, we separate the dataset based on four different regions and find the estimated regression functions for each region. We compare the slope for each estimated regression function and see whether the linear relationship between per capita income and the percentage of individuals in a county having at least a bachelor's degree is positive for all regions. Fourthly, we conduct analysis on MSE and conclude which region has a relatively higher variability around the fitted regression line. Lastly, we also find the 90% confidence interval for B1/ slope for all regions; prepare a residual plot and a normal probability plot to conclude which linear regression model is more appropriate.

Part I: Fitting regression models.

The estimated regression functions for Number of active physicians on Total population is  $\hat{Y} = -1.106348 \times 10^2 + 2.795425 \times 10^{-3}X_1$ .

The estimated regression functions for Number of active physicians on Number of hospital beds is  $\hat{Y} = -95.9321847 + 0.7431164X_2$ .

The estimated regression functions for Number of active physicians on Total personal income is  $\hat{Y} = -48.3948489 + 0.1317012X_3$ .

```
##      (Intercept) Total_population
##      -1.106348e+02      2.795425e-03

##
## Call:
## lm(formula = Number_of_active_physicians ~ Total_population,
##     data = CDI)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1969.4   -209.2    -88.0     27.9    3928.7
```

```
##
## Coefficients:
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -1.106e+02  3.475e+01  -3.184  0.00156 **
## Total_population  2.795e-03  4.837e-05  57.793  < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 610.1 on 438 degrees of freedom
## Multiple R-squared:  0.8841, Adjusted R-squared:  0.8838
## F-statistic: 3340 on 1 and 438 DF,  p-value: < 2.2e-16

##               (Intercept) Number_of_hospital_beds
##               -95.9321847              0.7431164

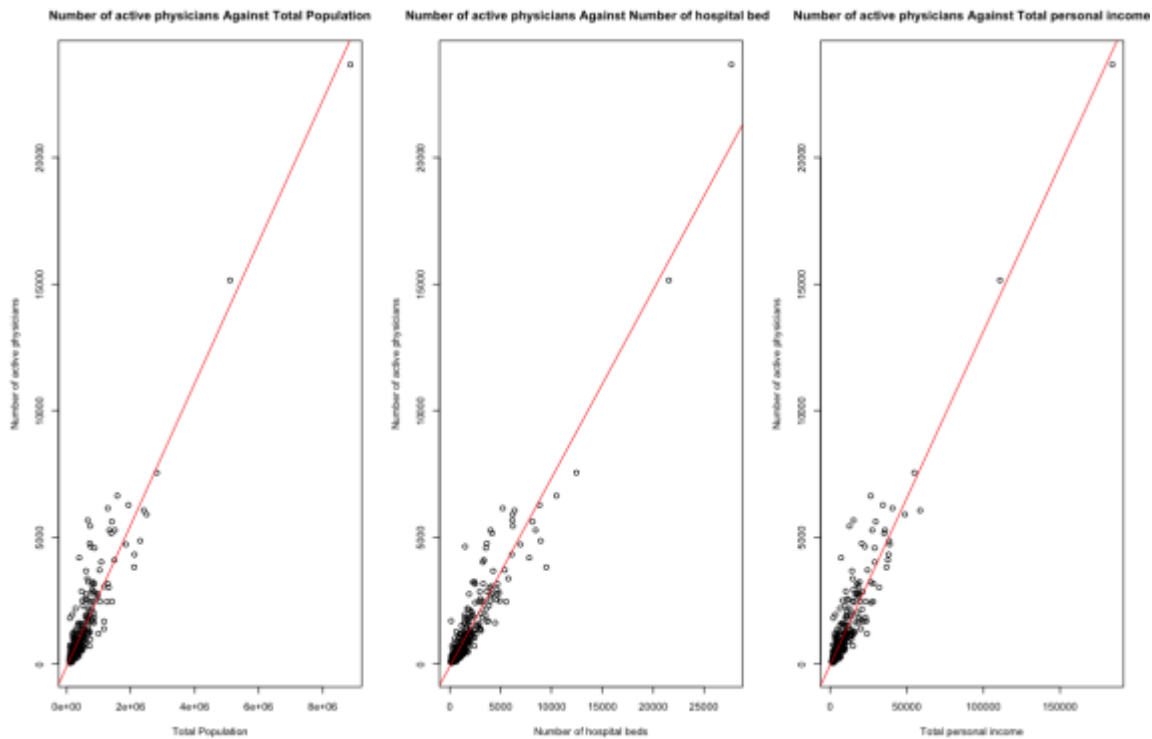
##
## Call:
## lm(formula = Number_of_active_physicians ~ Number_of_hospital_beds,
##     data = CDI)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3133.2  -216.8   -32.0    96.2   3611.1
##
## Coefficients:
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -95.93218   31.49396  -3.046  0.00246 **
## Number_of_hospital_beds  0.74312    0.01161  63.995  < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 556.9 on 438 degrees of freedom
## Multiple R-squared:  0.9034, Adjusted R-squared:  0.9032
## F-statistic: 4095 on 1 and 438 DF,  p-value: < 2.2e-16

##               (Intercept) Total_personal_income
##               -48.3948489              0.1317012

##
## Call:
## lm(formula = Number_of_active_physicians ~ Total_personal_income,
##     data = CDI)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1926.6  -194.5   -66.6    44.2   3819.0
##
## Coefficients:
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -48.39485   31.83333  -1.52   0.129
## Total_personal_income  0.13170    0.00211  62.41  <2e-16 ***
```

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 569.7 on 438 degrees of freedom
## Multiple R-squared:  0.8989, Adjusted R-squared:  0.8987
## F-statistic: 3895 on 1 and 438 DF, p-value: < 2.2e-16
```

As we can see from the graphs, the linear regression relation appears to provide a good fit for each of the three predictor variables even though there are a few outliers.



MSE for Total population is 372203.5.

MSE for Number of hospital beds is 310191.9.

MSE for Total personal income is 324539.4.

Number of hospital beds leads to the smallest variability around the fitted regression line because it has the smallest MSE.

```
## [1] 372203.5
```

```
## [1] 372203.5 310191.9 324539.4
```

The estimated regression functions for NE is  $Y_i = 9223.8156 + 522.1588X_i$

The estimated regression functions for NC is  $Y_i = 13581.4052 + 238.6694X_i$

The estimated regression functions for S is  $Y_i = 10529.7851 + 330.6117X_i$

The estimated regression functions for W is  $Y_i = 8615.0527 + 440.3157X_i$

```
##           (Intercept) Percent_bachelors_degrees
##           9223.8156             522.1588
```

```
##           (Intercept) Percent_bachelors_degrees
##           13581.4052             238.6694
```

```
##          (Intercept) Percent_bachelors_degrees
##          10529.7851          330.6117

##          (Intercept) Percent_bachelors_degrees
##          8615.0527          440.3157
```

As we can see from the regression functions, the directions of the relationship between per capita income in a county (Y) and the percentage of individuals in a county having at least a bachelor's degree (X) are the same for all regions because their slopes are all positive. Therefore, per capita income increases when the percentage of individuals in a county having at least a bachelor's degree increases. However, the increase in region NE is the highest and NC is the slowest because NE has the largest slope value and NC has the smallest slope value.

The MSE for NC is 4411341.

The MSE for NE is 7335008.

The MSE for S is 7474349.

The MSE for W is 8214318.

The variability around the fitted regression line is relatively higher in W because it has the largest MSE. It is approximately the same for NE and S because their MSE are relatively close. The MSE for NC is the smallest and therefore it has the smallest variability.

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

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

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

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

## Part II: Measuring linear associations.

The number of hospital beds accounts for the largest reduction in the variability in the number of active physicians because its  $R^2$  is closest to 1.

```
## [1] 0.8840674
```

```
## [1] 0.9033826
```

```
## [1] 0.8989137
```

## Part III. Inference about regression parameters

The 90 percent confidence coefficient for NC is (193.4858, 283.853).

The 90 percent confidence coefficient for NE is (460.5177, 583.8).

The 90 percent confidence coefficient for S is (285.7076, 375.5158).

The 90 percent confidence coefficient for W is (364.7585, 515.8729).

No, the regression lines for different regions do not appear to have similar slopes because their confidence intervals varies.

```
##          5 %    95 %
## Percent_bachelors_degrees 193.4858 283.853
```

```

##                                5 %   95 %
## Percent_bachelors_degrees 460.5177 583.8

##                                5 %   95 %
## Percent_bachelors_degrees 285.7076 375.5158

##                                5 %   95 %
## Percent_bachelors_degrees 364.7585 515.8729

## Analysis of Variance Table
##
## Response: Per_capita_income
##              Df      Sum Sq   Mean Sq F value    Pr(>F)
## Percent_bachelors_degrees  1 1450517671 1450517671  197.75 < 2.2e-16 ***
## Residuals                101  740835765    7335008
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

## Analysis of Variance Table
##
## Response: Per_capita_income
##              Df      Sum Sq   Mean Sq F value    Pr(>F)
## Percent_bachelors_degrees  1 338907694 338907694  76.826 3.344e-14 ***
## Residuals                106 467602149    4411341
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

## Analysis of Variance Table
##
## Response: Per_capita_income
##              Df      Sum Sq   Mean Sq F value    Pr(>F)
## Percent_bachelors_degrees  1 1109873245 1109873245  148.49 < 2.2e-16 ***
## Residuals                150 1121152411    7474349
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

## Analysis of Variance Table
##
## Response: Per_capita_income
##              Df      Sum Sq   Mean Sq F value    Pr(>F)
## Percent_bachelors_degrees  1 773745787 773745787  94.195 6.856e-15 ***
## Residuals                75 616073841    8214318
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

Part IV: Regression diagnostics.

The residuals for Total Population is

```

##           1           2           3           4           5
## -988.674369  992.803480 -214.428820 -967.381296 -565.893436
##           6           7           8           9          10

```

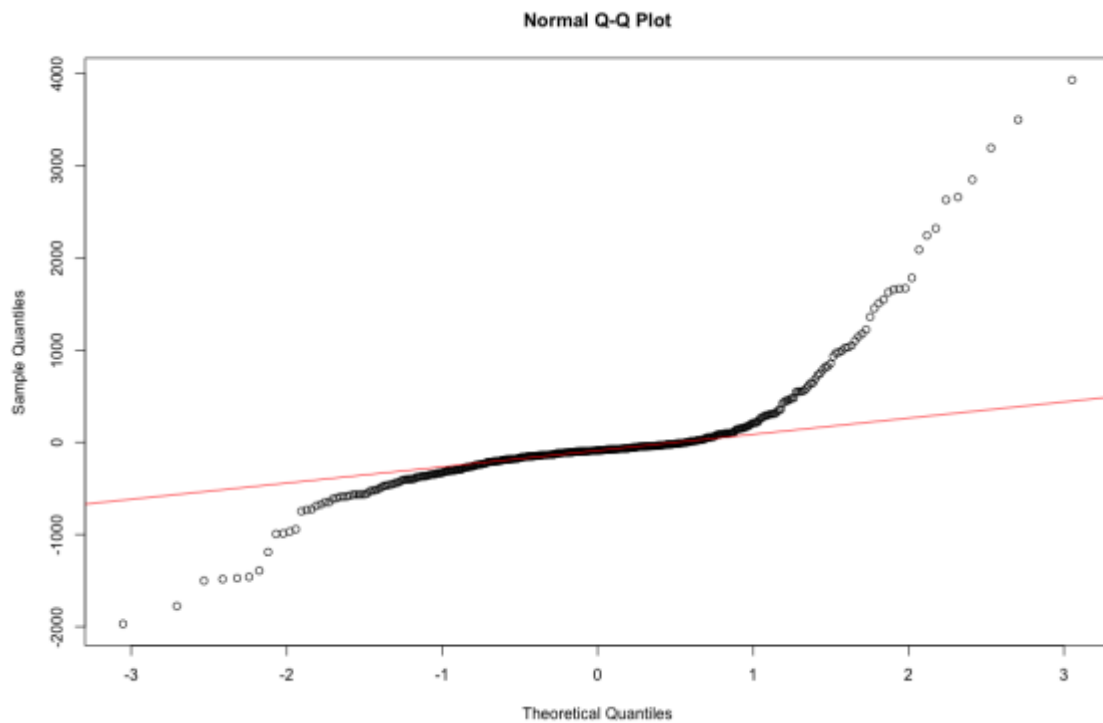
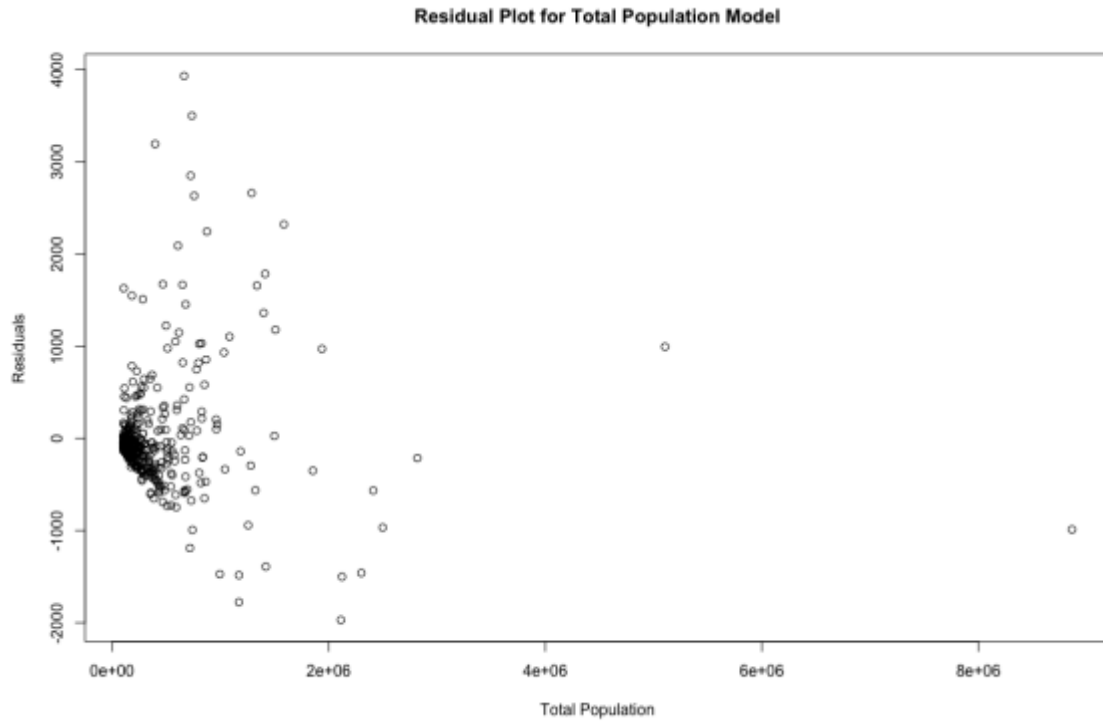
##	-1459.698605	-1501.539153	-1969.427598	969.634018	-350.756392
##	11	12	13	14	15
##	2319.273385	1177.037745	25.270774	-1391.339962	1783.103490
##	16	17	18	19	20
##	1359.322538	1655.691994	-563.536734	2658.950151	-296.222410
##	21	22	23	24	25
##	-942.987612	-141.045101	-1776.166841	-1483.300259	1101.534742
##	26	27	28	29	30
##	-336.014719	930.551475	-1472.700907	151.176329	98.009868
##	31	32	33	34	35
##	203.053577	2242.012596	853.159476	-470.264923	580.539388
##	36	37	38	39	40
##	-650.113979	-201.986164	-208.900524	214.015354	289.691337
##	41	42	43	44	45
##	1029.346991	-483.655301	1023.500977	-375.137651	818.236677
##	46	47	48	49	50
##	82.546194	746.218364	2629.422668	-993.058158	3497.162931
##	51	52	53	54	55
##	-674.979133	175.279314	2847.861778	-1189.803031	552.790868
##	56	57	58	59	60
##	28.707764	-555.173826	1452.026418	-416.240419	-232.605099
##	61	62	63	64	65
##	-130.275748	-577.523525	-591.549193	-567.283375	87.421834
##	66	67	68	69	70
##	422.853345	3928.735428	821.345583	108.662481	1664.541007
##	71	72	73	74	75
##	35.482291	1146.409850	2088.091419	356.467270	305.806785
##	76	77	78	79	80
##	-748.166535	-613.051673	1049.422531	-246.667691	-180.405177
##	81	82	83	84	85
##	-135.581813	-399.511928	-46.282453	-379.211552	-726.614349
##	86	87	88	89	90
##	-523.176486	-239.972948	-112.981081	-205.037755	975.776476
##	91	92	93	94	95
##	-732.758724	-138.417176	-39.971942	-281.841575	1221.481929
##	96	97	98	99	100
##	92.200859	264.715211	-566.924691	352.217875	-511.852522
##	101	102	103	104	105
##	333.655986	1670.466150	-689.032454	209.139335	-260.945549
##	106	107	108	109	110
##	-324.860418	94.429032	-519.002735	-86.792066	-563.963352
##	111	112	113	114	115
##	-530.146582	-587.351666	-400.681020	-467.679752	-311.892208
##	116	117	118	119	120
##	-497.332361	79.774118	549.496859	-463.580151	-87.772020
##	121	122	123	124	125
##	-441.678250	-451.188035	3190.731659	-274.742548	-309.160084
##	126	127	128	129	130
##	-400.804812	-134.610660	-650.536458	-110.920196	-97.833285
##	131	132	133	134	135

##	-347.551965	-430.662770	-367.131640	-402.569759	-47.576621
##	136	137	138	139	140
##	-321.921477	687.476236	-47.531139	-590.206872	-368.586035
##	141	142	143	144	145
##	291.107992	-604.898617	643.186514	-184.551734	-359.986555
##	146	147	148	149	150
##	-378.618830	-263.089989	-241.832048	156.073670	-317.148440
##	151	152	153	154	155
##	-212.681101	193.818262	-134.368020	-303.831554	-403.316946
##	156	157	158	159	160
##	-407.291025	-270.369300	-54.173115	-178.329661	642.875417
##	161	162	163	164	165
##	549.710230	-396.172871	-119.239199	-66.197268	-336.130943
##	166	167	168	169	170
##	310.925981	50.440339	1507.705648	-183.224466	51.570959
##	171	172	173	174	175
##	-367.245814	-206.869450	-34.867164	-191.706047	-457.741626
##	176	177	178	179	180
##	-365.893850	-437.983813	-356.277588	561.392801	299.519358
##	181	182	183	184	185
##	-191.259042	-28.610503	-345.458026	-335.211266	-295.148243
##	186	187	188	189	190
##	-101.997290	481.388732	-346.511142	-79.671753	-116.078363
##	191	192	193	194	195
##	1.732310	475.364076	-75.364515	-153.558418	-311.091329
##	196	197	198	199	200
##	314.662165	-121.128687	-162.980530	-233.606452	162.969149
##	201	202	203	204	205
##	221.952883	-109.581552	-158.225771	-298.824501	269.846401
##	206	207	208	209	210
##	468.418694	-333.542170	-102.569362	-152.072285	-301.199351
##	211	212	213	214	215
##	-148.367585	-288.512694	728.641054	-67.283469	-137.454499
##	216	217	218	219	220
##	100.106110	143.272564	-267.472037	-72.672293	452.912204
##	221	222	223	224	225
##	-72.425280	-1.790719	-296.299233	-258.776489	-97.582589
##	226	227	228	229	230
##	-312.530747	-22.442565	-146.844344	-159.176237	-54.846633
##	231	232	233	234	235
##	19.328969	99.251459	-97.593777	-225.264426	-37.454774
##	236	237	238	239	240
##	-60.191498	-199.006238	-144.338132	-14.464944	-269.267740
##	241	242	243	244	245
##	89.462628	-186.743980	612.289565	-117.044362	-186.205735
##	246	247	248	249	250
##	-248.740170	-251.256561	281.973426	-75.005482	-131.694428
##	251	252	253	254	255
##	-212.574225	-241.601673	180.555633	-273.694431	-151.501547
##	256	257	258	259	260

##	-71.468002	-207.277913	1546.328694	782.996801	-181.108663
##	261	262	263	264	265
##	-30.523404	-203.077407	211.970115	-236.824551	-144.659621
##	266	267	268	269	270
##	90.935805	-159.476394	-98.026331	8.956388	-43.403460
##	271	272	273	274	275
##	-307.445138	247.796030	-10.713246	-111.371189	-212.909182
##	276	277	278	279	280
##	-38.630658	-93.460137	60.227538	11.054983	-177.455817
##	281	282	283	284	285
##	-44.100545	151.795515	-141.485299	-57.093939	-75.806011
##	286	287	288	289	290
##	111.638462	-151.790256	-65.964844	-162.004998	-108.059891
##	291	292	293	294	295
##	-210.214149	-168.905128	-147.560528	-32.546551	-187.306145
##	296	297	298	299	300
##	-102.202714	36.661834	-11.138166	-137.260403	-227.201699
##	301	302	303	304	305
##	-133.156972	129.413294	-204.226096	-180.996871	96.319774
##	306	307	308	309	310
##	-96.109197	-128.824573	-33.393315	-31.177305	48.914944
##	311	312	313	314	315
##	-35.584675	-184.206531	-153.398653	-171.683024	-96.249733
##	316	317	318	319	320
##	-11.326481	-117.054563	-113.280230	86.222946	50.341116
##	321	322	323	324	325
##	-186.950879	-18.849482	-23.402214	-122.022036	55.222690
##	326	327	328	329	330
##	-54.150372	-39.619242	-191.469560	-108.354948	96.086730
##	331	332	333	334	335
##	-94.282013	-137.586462	-97.492689	-97.173248	-100.373504
##	336	337	338	339	340
##	-179.833987	-154.824838	-85.860926	438.306800	-11.120138
##	341	342	343	344	345
##	-135.955208	-122.406543	-114.640596	-22.442121	-41.285577
##	346	347	348	349	350
##	-79.478209	-79.784943	-85.133609	-147.018235	-107.685579
##	351	352	353	354	355
##	-43.564614	-66.367410	-165.481260	-97.971731	-94.574780
##	356	357	358	359	360
##	140.536018	-57.399687	-137.136917	-92.242381	25.654951
##	361	362	363	364	365
##	11.536271	-106.396638	-60.021289	-150.766906	-91.840858
##	366	367	368	369	370
##	-77.620019	-99.443908	-43.702358	71.389891	12.169815
##	371	372	373	374	375
##	-6.862968	-19.851787	-15.350643	-68.836285	-22.436539
##	376	377	378	379	380
##	10.789890	-87.818751	-41.377073	-46.005791	-153.693975
##	381	382	383	384	385



##	-80.729553	27.633852	-144.642133	-39.957254	-41.518372
##	386	387	388	389	390
##	1.994715	-45.328030	55.580483	1.608437	-78.000203
##	391	392	393	394	395
##	-62.018247	542.487725	-70.380128	-100.678476	-74.015961
##	396	397	398	399	400
##	451.831053	-76.637816	-16.587499	-75.498045	145.999541
##	401	402	403	404	405
##	-84.553191	-34.878732	-28.177080	-112.913548	-22.005035
##	406	407	408	409	410
##	63.112373	-112.042647	-59.866535	304.941343	-6.644934
##	411	412	413	414	415
##	-92.510754	-76.471618	-39.660183	-84.232483	-70.182165
##	416	417	418	419	420
##	-48.388264	-118.206562	1627.005890	-6.979370	-21.641124
##	421	422	423	424	425
##	14.391150	-100.385216	-16.739473	-68.951163	-41.740744
##	426	427	428	429	430
##	-65.745573	-92.582676	-82.079500	166.058238	-88.279246
##	431	432	433	434	435
##	-137.128293	-52.966158	-91.014190	-88.991826	-105.133631
##	436	437	438	439	440
##	-74.024609	21.576407	-83.299832	22.046800	-47.027914



As we can see from the residual plot for total population model, there are two outliers (5e+06, 1000), (9e+06, -1000). The residuals are not centered around the 0 line, which indicates that it is not a good fit. In addition, the qqline is not linear, which means that the data is not normal.

1. The normality probability plot shows departures from normality in the tails.

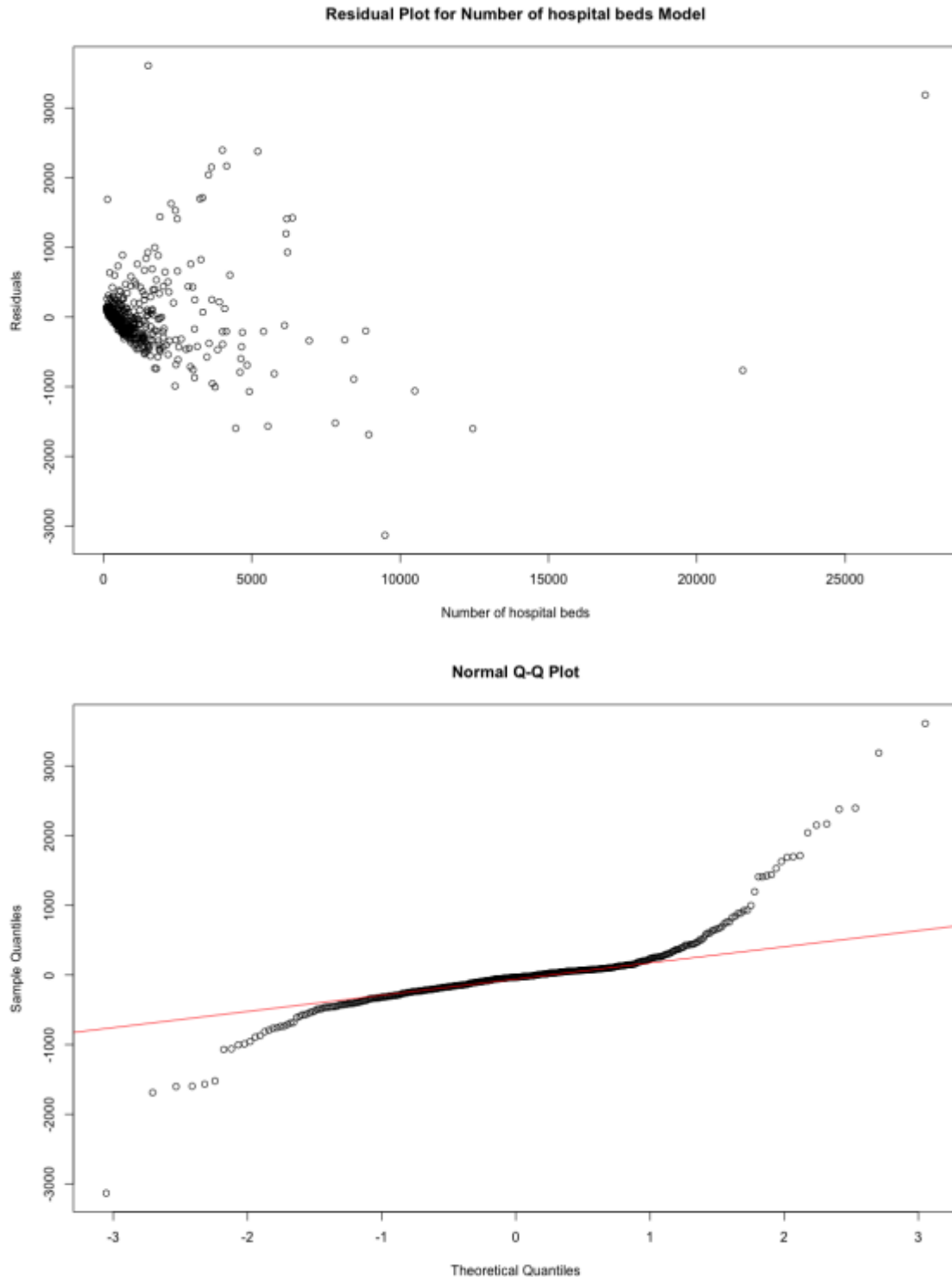
The residuals for Number of hospital beds is

##	1	2	3	4	5
##	3188.6066863	-765.2271832	-1602.1244265	1409.2156773	1425.0235530
##	6	7	8	9	10
##	-1688.0150574	-120.0505894	-3133.2428687	-199.2171801	-338.8372379
##	11	12	13	14	15
##	-1061.3317785	2396.7783608	1713.4370289	70.2352138	-327.0907378
##	16	17	18	19	20
##	2168.5127093	-891.9981367	215.8055874	2378.7266760	824.5377827
##	21	22	23	24	25
##	-1567.1622643	121.5583946	-328.5563564	-955.7913976	1697.8312760
##	26	27	28	29	30
##	438.3347372	-207.1810306	505.2012199	-598.1831838	-209.7078721
##	31	32	33	34	35
##	-207.3130098	2042.2999730	-220.0821225	-422.2882440	762.1698394
##	36	37	38	39	40
##	-1596.8809226	505.3970355	-171.9490654	659.5997734	-1069.7144868
##	41	42	43	44	45
##	1628.3698089	1689.6114648	1409.5447050	533.4417980	-426.5317456
##	46	47	48	49	50
##	884.1116949	-690.4945206	3611.0557037	-429.4749841	930.3808827
##	51	52	53	54	55
##	366.5322458	647.2224470	2151.9883286	-92.2937809	247.3628866
##	56	57	58	59	60
##	359.3604257	888.5945441	1530.8748081	-713.6558797	-315.2622478
##	61	62	63	64	65
##	335.8732700	-211.9887512	244.3764236	34.7534424	438.8094337
##	66	67	68	69	70
##	-377.8192394	1196.7935884	-791.8896905	689.7349837	-814.1891833
##	71	72	73	74	75
##	-1004.6718775	1439.7815918	602.7699005	-573.3148554	204.1773769
##	76	77	78	79	80
##	-335.2099478	99.8714244	430.8948048	315.5140947	-870.9490654
##	81	82	83	84	85
##	-18.9615246	-683.5012879	289.8632717	446.7788234	-109.3488494
##	86	87	88	89	90
##	-91.4399124	75.0006352	-342.2450196	-291.9061486	-469.8367753
##	91	92	93	94	95
##	-302.5778911	-612.7763226	-165.7684774	17.9821765	-389.9096872
##	96	97	98	99	100
##	-430.3266994	760.7794386	-19.0831979	398.3132024	-990.2077946
##	101	102	103	104	105
##	390.4139562	249.0709313	98.5579344	-758.0225925	-237.3566945
##	106	107	108	109	110
##	-493.9064562	108.2762850	-12.5512797	28.1749161	425.4834844
##	111	112	113	114	115

##	67.7785158	-487.1645701	-571.9709078	253.4293388	-34.0644316
##	116	117	118	119	120
##	-252.2753222	54.6889908	-447.9312219	-302.2017951	-341.4020722
##	121	122	123	124	125
##	-147.3491570	-118.5052868	-1521.7797086	-39.6517258	381.0804694
##	126	127	128	129	130
##	127.5397833	342.9252624	-232.1479571	-457.1542642	-464.5028259
##	131	132	133	134	135
##	6.2650563	-411.5318982	-316.5778911	-352.5778911	204.7791310
##	136	137	138	139	140
##	-451.2377897	671.9452591	-52.3025489	-160.4326825	163.5760855
##	141	142	143	144	145
##	581.5494741	40.7040658	-463.6471118	-110.8901507	203.1446134
##	146	147	148	149	150
##	-150.7715535	-110.6520335	-97.3400814	-538.5754303	-288.2937809
##	151	152	153	154	155
##	-421.0268990	-191.2005647	-263.2199462	-369.6238840	-33.5522025
##	156	157	158	159	160
##	230.4559502	-157.7806290	279.8423522	-294.3394662	-0.6035797
##	161	162	163	164	165
##	427.0090955	-248.5240531	-364.8895355	-236.8347747	191.3551965
##	166	167	168	169	170
##	197.9906368	-334.9899817	998.3407366	106.9340304	-243.6692617
##	171	172	173	174	175
##	-295.3403891	301.5570116	-285.9355284	-489.0638164	86.6302311
##	176	177	178	179	180
##	-48.5340514	-28.4151466	-312.3403891	94.2578264	473.3104340
##	181	182	183	184	185
##	298.6211556	-397.6783373	-46.3873048	-17.3231608	-383.5785063
##	186	187	188	189	190
##	-559.1642625	103.8078956	-247.1388816	80.2828997	199.2272161
##	191	192	193	194	195
##	-36.7530948	-401.9884436	-59.3037793	-139.5881971	-461.3582325
##	196	197	198	199	200
##	-394.2093326	-138.3954575	-175.9367588	56.9703325	7.5316306
##	201	202	203	204	205
##	601.0066347	-736.2005647	-198.3954575	129.7403680	-139.0087479
##	206	207	208	209	210
##	734.2913601	43.0157102	235.1167716	101.8601956	-366.4505259
##	211	212	213	214	215
##	-147.3406967	103.2541351	-31.5763531	260.3461209	-260.7530948
##	216	217	218	219	220
##	120.6408447	-409.8157008	-328.6707997	-319.2480956	-137.1091941
##	221	222	223	224	225
##	132.0254010	-236.4133010	178.2353688	62.1533813	-93.2124087
##	226	227	228	229	230
##	-187.5981954	-50.4508336	-455.6057329	-217.4323749	-303.9821365
##	231	232	233	234	235
##	-740.5306678	-332.3300831	49.0163254	-178.5888123	-531.7427888
##	236	237	238	239	240

##	-73.7721687	-4.2133315	-84.7174079	-102.6526487	15.6211556
##	241	242	243	244	245
##	183.8695788	147.6117724	-200.0169007	-19.2130239	-58.0021332
##	246	247	248	249	250
##	77.6393066	26.3367378	636.3639644	146.5754702	69.9975591
##	251	252	253	254	255
##	-164.8275448	-19.7640160	-228.9267605	-31.7274062	95.4744089
##	256	257	258	259	260
##	-41.5155928	29.8138951	928.2299845	461.4765621	146.7034506
##	261	262	263	264	265
##	-412.0278218	30.5019431	-292.4126858	-35.5346666	-65.4057635
##	266	267	268	269	270
##	26.1727628	75.8780391	68.5113263	-120.2765527	185.6486898
##	271	272	273	274	275
##	-13.9845974	130.5582420	-171.3131624	-284.0012104	48.4650257
##	276	277	278	279	280
##	-158.8087785	47.0250934	-515.5960422	-299.5603552	-32.2868586
##	281	282	283	284	285
##	-265.9276833	-167.3034717	-216.3319287	7.3739627	93.1533813
##	286	287	288	289	290
##	55.7322153	-212.5246683	-63.8735377	-95.2774755	-148.8275448
##	291	292	293	294	295
##	-311.0475109	-228.2036407	-230.4971341	57.6489974	28.9787929
##	296	297	298	299	300
##	151.8777315	-145.8453882	-148.2859358	88.8411217	65.7216017
##	301	302	303	304	305
##	-43.8923040	-216.6154237	101.0976977	-202.8735377	-36.8547714
##	306	307	308	309	310
##	-266.5243607	-75.1307288	15.1536890	-80.4514488	268.5567040
##	311	312	313	314	315
##	-28.9745990	140.5654719	-81.3510026	-84.0024409	-43.9289137
##	316	317	318	319	320
##	-2.5434346	-104.3416195	-84.2499413	-256.8081633	261.2998204
##	321	322	323	324	325
##	-14.8560018	-411.1382663	-185.7721687	-225.3228532	310.0886221
##	326	327	328	329	330
##	-55.8003181	-54.4332977	46.4556426	-97.9379893	26.5485512
##	331	332	333	334	335
##	-119.8919963	73.8777315	-152.6717226	88.5842382	72.4099573
##	336	337	338	339	340
##	43.6849919	-80.9382969	59.1164640	366.1264623	-228.7812442
##	341	342	343	344	345
##	11.8320462	34.9787929	-250.8090861	-143.9467572	-177.0659696
##	346	347	348	349	350
##	-170.0202843	-15.7089475	131.3730399	54.2538275	-12.2687076
##	351	352	353	354	355
##	139.9052657	-81.2774755	9.7034506	-104.1307288	121.8042044
##	356	357	358	359	360
##	-205.7900122	73.2907449	-26.7733991	146.5288621	-57.4974417
##	361	362	363	364	365

##	-0.6720302	52.9878684	58.6026969	-24.9661387	11.9697173
##	366	367	368	369	370
##	-239.3228532	81.9969439	-12.6263449	-18.7543252	59.4653334
##	371	372	373	374	375
##	-54.1673386	-47.1031946	71.8229707	-40.1401120	-196.2677847
##	376	377	378	379	380
##	-168.4695998	69.1621493	-69.8738453	-271.8272372	-72.6907965
##	381	382	383	384	385
##	-132.2224070	-32.3966879	94.8223554	79.1806080	8.8138951
##	386	387	388	389	390
##	-236.2859358	112.2813617	-53.7543252	-179.8366203	131.7857457
##	391	392	393	394	395
##	31.3458133	81.7240625	-52.9658311	-111.4611395	-34.9292213
##	396	397	398	399	400
##	255.9890988	75.5107110	-190.7174079	-29.7549404	-316.3034717
##	401	402	403	404	405
##	28.6117724	78.8411217	-152.2130239	13.7034506	56.1439982
##	406	407	408	409	410
##	262.2719786	-184.4792906	-24.7274062	158.6952979	144.6849919
##	411	412	413	414	415
##	-22.3147005	91.7766701	-140.7452497	-174.5985030	67.5382453
##	416	417	418	419	420
##	-12.5255911	57.1803004	842.0738547	-153.1028870	54.0889298
##	421	422	423	424	425
##	-97.6901812	-29.8835360	-50.1767217	13.4925600	58.8135875
##	426	427	428	429	430
##	-102.9198382	37.0244782	-0.5349743	-38.8916887	24.9052657
##	431	432	433	434	435
##	16.0335537	30.1806080	44.1712248	64.7491359	85.6480746
##	436	437	438	439	440
##	-21.5715840	83.0889298	43.2262933	152.6849919	-126.8738453



As we can see from the residual plot for number of hospital beds model, there are two outliers (22000, -1000), (28000, 3000). The residuals are not mainly centered around the 0 line, which indicates that it is not a good fit. In addition, the qqline is not linear (more of a cubic function line), which m

means that the data is not normal. The normality probability plot shows departures from normality in the tails.

The residuals for Total Personal Income is

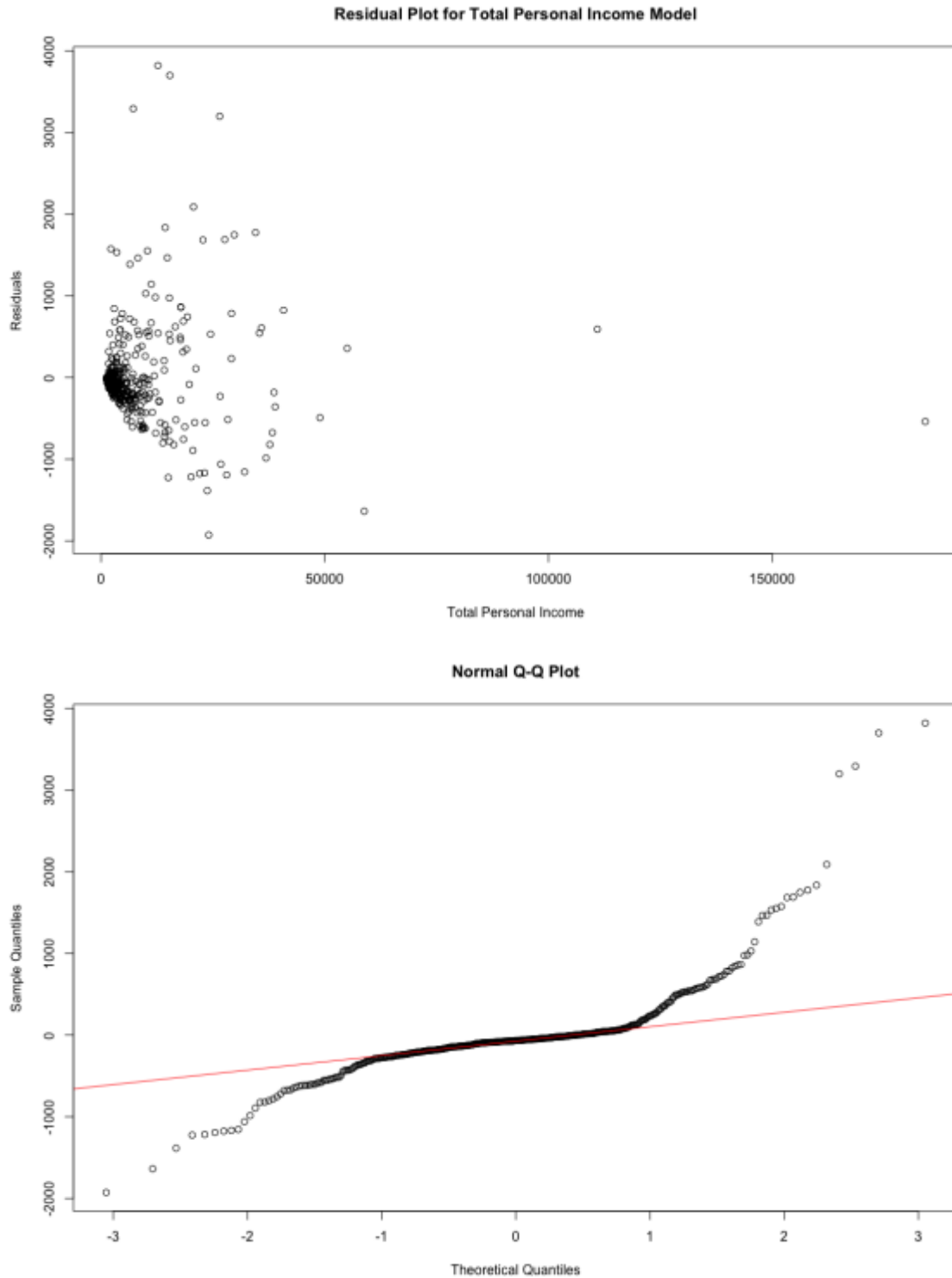
##	1	2	3	4	5
##	-5.379152e+02	5.920453e+02	3.574343e+02	-4.908760e+02	-1.636006e+03
##	6	7	8	9	10
##	-1.819097e+02	-6.740486e+02	-9.846914e+02	1.775411e+03	-3.582301e+02
##	11	12	13	14	15
##	3.197733e+03	6.078291e+02	-8.194276e+02	-5.519748e+02	1.746860e+03
##	16	17	18	19	20
##	5.444362e+02	1.689306e+03	-1.154394e+03	8.243570e+02	-5.138315e+02
##	21	22	23	24	25
##	-1.191931e+03	6.893319e+02	-1.215643e+03	-1.175587e+03	2.318075e+02
##	26	27	28	29	30
##	-8.396239e+01	5.311399e+02	-1.926599e+03	4.614645e+02	3.124729e+02
##	31	32	33	34	35
##	4.939178e+02	7.851199e+02	8.645577e+02	-1.166302e+03	1.089101e+02
##	36	37	38	39	40
##	-7.554338e+02	-2.732569e+02	2.067006e+02	-1.059983e+03	5.317174e+02
##	41	42	43	44	45
##	-2.291202e+02	-1.383928e+03	8.583969e+02	-8.923233e+02	9.741079e+02
##	46	47	48	49	50
##	-5.507159e+02	3.460094e+02	1.684295e+03	-8.013986e+02	3.819000e+03
##	51	52	53	54	55
##	-5.779811e+02	6.701073e+02	2.088975e+03	-1.223572e+03	4.514798e+02
##	56	57	58	59	60
##	9.077406e+01	-8.219593e+02	7.435619e+02	-1.799017e+02	2.099451e+01
##	61	62	63	64	65
##	-5.173075e+02	-7.817650e+02	-6.708832e+02	-2.021772e+02	5.551301e+02
##	66	67	68	69	70
##	5.432338e+02	3.698279e+03	9.823326e+02	-6.031831e+02	1.466164e+03
##	71	72	73	74	75
##	1.912963e+02	6.230525e+02	1.835775e+03	5.693576e+02	5.073821e+02
##	76	77	78	79	80
##	-5.612427e+01	-2.868445e+02	1.141839e+03	-8.981842e+01	-2.989651e+01
##	81	82	83	84	85
##	-6.470099e+02	-2.605377e+02	-2.839258e+02	-4.255593e+02	-4.063181e+02
##	86	87	88	89	90
##	-6.796231e+02	-7.205966e+02	4.014468e+00	-2.481249e+02	1.029256e+03
##	91	92	93	94	95
##	-6.146989e+02	-7.268710e+01	-1.638277e+01	-1.758796e+02	1.463440e+03
##	96	97	98	99	100
##	-2.967160e+02	2.615577e+02	-2.637499e+02	5.246025e+02	-2.672649e+02
##	101	102	103	104	105
##	3.840457e+02	1.550971e+03	-5.868449e+02	3.545722e+02	-1.938006e+02
##	106	107	108	109	110
##	-4.316905e+02	2.757735e+02	-6.229108e+02	-2.105985e+02	-6.130463e+02
##	111	112	113	114	115



##	-6.080649e+02	-6.352422e+02	-2.136486e+02	-6.183747e+02	-3.424638e+02
##	116	117	118	119	120
##	-4.288440e+02	-5.537286e+02	5.748693e+02	-1.846561e+02	-4.637210e+01
##	121	122	123	124	125
##	-3.169668e+02	-2.824472e+02	3.291122e+03	-4.297690e+02	-1.887186e+02
##	126	127	128	129	130
##	-5.291560e+02	-2.392307e+02	-9.010131e+01	-5.778134e+00	1.808868e+02
##	131	132	133	134	135
##	-5.836115e+02	-2.206037e+02	-1.371901e+02	-2.620884e+02	-1.427263e+02
##	136	137	138	139	140
##	-1.746012e+02	6.802303e+02	1.215140e+02	-5.377152e+02	-3.483725e+02
##	141	142	143	144	145
##	-1.714095e+00	-6.036988e+02	7.190632e+02	-2.067326e+02	-2.447447e+02
##	146	147	148	149	150
##	-3.637304e+02	-2.556768e+02	-3.640855e+02	2.637867e+02	-2.757469e+02
##	151	152	153	154	155
##	-6.655265e+01	-7.443124e+01	-5.533106e+01	-3.852361e+02	-1.978502e+02
##	156	157	158	159	160
##	-4.164695e+02	-2.744020e+02	-1.878305e+02	-1.572821e+02	7.831067e+02
##	161	162	163	164	165
##	4.960910e+02	-2.865704e+02	5.251451e+01	-7.320400e+01	-2.718127e+02
##	166	167	168	169	170
##	4.025858e+02	1.003634e+02	1.387449e+03	-3.962183e+01	1.291616e+02
##	171	172	173	174	175
##	-1.598738e+02	-2.219955e+02	-2.332944e+02	-7.626050e+01	-5.163159e+02
##	176	177	178	179	180
##	-1.278810e+02	-3.768582e+02	-2.321883e+02	5.194870e+02	4.392130e+01
##	181	182	183	184	185
##	-7.510093e+01	8.242123e+01	-4.451935e+02	-5.873785e+01	-1.831254e+02
##	186	187	188	189	190
##	3.766354e+01	5.847919e+02	-1.480916e+01	-1.742821e+02	-1.458418e+02
##	191	192	193	194	195
##	-6.387431e+01	5.809957e+02	-4.275228e+01	-5.892617e+01	-2.667582e+02
##	196	197	198	199	200
##	4.122148e+02	-8.934324e+01	-9.459861e+01	-3.134699e+02	9.633887e+01
##	201	202	203	204	205
##	-2.185270e+02	-5.179911e+01	-1.501334e+02	-2.220148e+02	2.977826e+02
##	206	207	208	209	210
##	-8.824002e+01	-3.139156e+02	-1.890132e+02	-5.776320e+01	-2.110021e+02
##	211	212	213	214	215
##	-4.783201e+01	-2.638726e+02	7.229679e+02	-1.353268e+02	-2.827380e+02
##	216	217	218	219	220
##	1.242638e+02	2.506062e+02	-1.966070e+02	-1.200677e+01	4.902917e+02
##	221	222	223	224	225
##	-2.016760e+01	3.455506e+01	-3.391731e+02	-2.775450e+02	-6.385988e+01
##	226	227	228	229	230
##	-3.054880e+02	3.847593e+00	-8.849051e+01	-1.313867e+02	-1.370960e+01
##	231	232	233	234	235
##	8.297214e+01	1.473533e+02	-9.376195e+01	-1.635209e+02	5.191601e+01
##	236	237	238	239	240

##	-7.166992e+01	-7.074965e+01	-1.364660e+02	4.508739e+01	-2.487889e+02
##	241	242	243	244	245
##	2.206839e+02	-1.436049e+02	6.821305e+02	-2.239761e+01	-1.029337e+02
##	246	247	248	249	250
##	-1.170970e+02	-2.425792e+02	6.381671e+01	-1.874517e+02	-9.373194e+01
##	251	252	253	254	255
##	-1.262659e+02	-3.190684e+02	1.878008e+02	-2.365910e+02	-2.069135e+02
##	256	257	258	259	260
##	1.440189e+01	-1.483335e+02	1.531704e+03	8.443542e+02	-1.867598e+02
##	261	262	263	264	265
##	4.226554e+01	-9.467281e+01	1.878358e+02	-1.858754e+02	-1.423415e+02
##	266	267	268	269	270
##	1.275593e+02	-6.609704e+01	-2.504564e+02	4.398356e+01	-8.158380e+01
##	271	272	273	274	275
##	-2.493592e+02	-1.736860e+01	5.918155e+01	-3.509704e+01	-2.067134e+02
##	276	277	278	279	280
##	6.322336e+01	-1.034593e+02	6.613510e+01	4.339724e+01	-1.677286e+02
##	281	282	283	284	285
##	3.858004e+01	2.407991e+02	-9.753057e+01	-3.618317e+01	-1.139219e+02
##	286	287	288	289	290
##	8.479951e+01	-7.883954e+01	-4.717132e+00	-1.325306e+02	-1.001249e+02
##	291	292	293	294	295
##	-2.152659e+02	-1.231903e+02	-1.350620e+02	-1.821862e+02	-1.699257e+02
##	296	297	298	299	300
##	-7.834061e+01	4.838670e+01	4.777464e+01	-1.743136e+02	-1.906927e+02
##	301	302	303	304	305
##	-5.520678e+01	1.397573e+02	-4.871287e+01	-1.352883e+02	4.483328e+01
##	306	307	308	309	310
##	-9.161330e+01	-1.027218e+02	2.339032e-01	1.677464e+01	5.173284e+01
##	311	312	313	314	315
##	1.098444e+01	-2.359476e+02	-1.437847e+02	-9.653898e+01	-2.430470e+01
##	316	317	318	319	320
##	4.434406e+00	-5.541657e+01	-7.283025e+01	1.276185e+02	-6.676446e+01
##	321	322	323	324	325
##	-1.516741e+02	5.668010e+01	-5.346511e+00	-7.155291e+01	2.087044e+01
##	326	327	328	329	330
##	-4.153182e+01	-3.569730e+01	-1.380704e+02	-8.825913e+01	5.502662e+01
##	331	332	333	334	335
##	-7.641531e+01	-1.446192e+02	-6.513333e+01	1.631162e+01	-4.845160e+01
##	336	337	338	339	340
##	-1.725715e+02	-2.119773e+00	-5.525574e+01	3.984158e+02	4.113346e+01
##	341	342	343	344	345
##	-1.200844e+02	-2.395222e+02	-8.725574e+01	7.773387e+00	7.343255e+00
##	346	347	348	349	350
##	-1.031414e+02	-7.990834e+01	-1.595293e+02	-1.279679e+02	-1.042452e+02
##	351	352	353	354	355
##	-1.793408e+01	-2.133509e+01	-1.251544e+02	-7.653898e+01	-1.239805e+02
##	356	357	358	359	360
##	1.222723e+02	-2.594463e+01	-1.469037e+02	-1.670422e+02	7.731287e+01
##	361	362	363	364	365

##	1.548086e+01	-1.310983e+02	1.557502e+01	-1.104575e+02	-7.518819e+01
##	366	367	368	369	370
##	-4.340853e+01	-1.087382e+02	-3.471964e+01	7.422210e+01	-4.017049e+01
##	371	372	373	374	375
##	1.806002e+01	-7.633075e-02	2.265163e+00	-2.987244e+01	1.652266e+01
##	376	377	378	379	380
##	3.371852e+01	-6.067194e+01	-4.390960e+01	-5.875806e+01	-6.524093e+01
##	381	382	383	384	385
##	-2.906365e+01	2.783629e+01	-2.057463e+02	-2.746679e+01	1.386290e+01
##	386	387	388	389	390
##	3.405411e+01	-3.680828e+01	5.239222e+01	3.632216e+01	-8.503201e+01
##	391	392	393	394	395
##	-6.655417e+01	5.402116e+02	-7.015943e+00	-9.333044e+01	-1.886441e+01
##	396	397	398	399	400
##	2.362799e+02	-4.709742e+01	-7.774511e+00	-4.978506e+01	6.269316e+01
##	401	402	403	404	405
##	-6.623842e+01	-7.144570e+01	-2.277451e+01	-7.861494e+01	1.559950e+01
##	406	407	408	409	410
##	-6.310508e+01	-5.124557e+01	-6.628148e+01	3.190541e+02	-1.970827e+02
##	411	412	413	414	415
##	-8.697966e+01	-3.173270e+01	-3.742837e+01	-7.452630e+01	4.665789e+00
##	416	417	418	419	420
##	-8.417300e+01	-8.911262e+01	1.574759e+03	1.733145e+01	-1.136547e+01
##	421	422	423	424	425
##	2.250283e+01	-1.887621e+01	-1.753095e+01	-2.259171e+01	-4.848198e+01
##	426	427	428	429	430
##	-8.467658e+01	-8.831186e+01	-8.028738e+01	1.709271e+02	-2.309529e+01
##	431	432	433	434	435
##	-1.305706e+02	-7.264745e+01	-5.543553e+01	-8.754614e+01	-1.419493e+02
##	436	437	438	439	440
##	-3.890872e+01	-1.190713e+02	-3.884582e+01	-4.174259e+00	-4.651701e+01



As we can see from the residual plot for total personal income model, there are two outliers (110000, 800), (180000, -500). The residuals are not mainly centered around the 0 line, which indicates that it is not a good fit. In addition, the qqline is not linear (more of a cubic function line), which

means that the data is not normal. The normality probability plot shows departures from normality in the tails.

Since these three residual plots and normal Q-Q plots are relatively similar, we can not conclude which linear regression model is more appropriate in one case than in the others.

#### Part V: Discussion.

In this report, we find that the number of active physicians has positive linear relationship with total population, number of hospital bed and total personal income. But we can not conclude that the increase in the number of active physicians is caused by the increase in total population, number of hospital bed and total personal income since there are many other factors that can affect the results. In other words, correlation does not prove causation. We also conclude that per capita income and the percentage of individuals in a county having at least a bachelor's degree has a linear relationship because 0 is not within the confidence intervals for all B1 in four different regions. As we can see from the dataset, there are seventeen different variables. However, we only make analysis on a few variables in this report. Therefore, our results might be biased. In addition, we should use transformations in y variables because they have unequal variance and nonnormality.

#### Appendix:

```
CDI <- read.table("~/Desktop/CDI.txt", quote="\"", comment.char="")
Identification_number = CDI$V1
County = CDI$V2
State = CDI$V3
Land_area = CDI$V4
Total_population = CDI$V5
Percent_of_population_aged_18_34 = CDI$V6
Percent_of_population_65_older = CDI$V7
Number_of_active_physicians = CDI$V8
Number_of_hospital_beds = CDI$V9
Total_serious_crimes = CDI$V10
Percent_high_school_graduates = CDI$V11
Percent_bachelors_degrees = CDI$V12
Percent_below_poverty_level = CDI$V13
Percent_unemployment = CDI$V14
Per_capita_income = CDI$V15
Total_personal_income = CDI$V16
Geographic_region = CDI$V17

names(CDI) = c('Identification_number', 'County', 'State', 'Land_area', 'Total_population',
               'Percent_of_population_aged_18_34', 'Percent_of_population_65_older',
               'Number_of_active_physicians', 'Number_of_hospital_beds', 'Total_serious_crimes',
               'Percent_high_school_graduates', 'Percent_bachelors_degrees',
               'Percent_below_poverty_level', 'Percent_unemployment', 'Per_capita_income',
               'Total_personal_income', 'Geographic_region' )
```

1.43

a.

```
# Regress Y (Number_of_active_physicians) on Total_population.
fit1 = lm(Number_of_active_physicians~Total_population, data = CDI)
fit1$coefficients
summary(fit1)
# Regress Y (Number_of_active_physicians) on Number_of_hospital_beds.
fit2 = lm(Number_of_active_physicians~Number_of_hospital_beds, data = CDI)
fit2$coefficients
summary(fit2)
# Regress Y (Number_of_active_physicians) on Total_personal_income.
fit3 = lm(Number_of_active_physicians~Total_personal_income, data = CDI)
fit3$coefficients
summary(fit3)
```

b.

```
par(mfrow=c(1,3))
plot(CDI$Total_population, CDI$Number_of_active_physicians, xlab = 'Total Pop
ulation', ylab = 'Number of active physicians', main = 'Number of active phys
icians Against Total Population')
abline(fit1, col = 'red')

plot(CDI$Number_of_hospital_beds, CDI$Number_of_active_physicians, xlab = 'Nu
mber of hospital beds', ylab = 'Number of active physicians', main = 'Number
of active physicians Against Number of hospital beds')
abline(fit2, col = 'red')

plot(CDI$Total_personal_income, CDI$Number_of_active_physicians, xlab = 'Tota
l personal income', ylab = 'Number of active physicians', main = 'Number of a
ctive physicians Against Total personal income')
abline(fit3, col = 'red')
```

c.

```
# MSE 1
anova.1 = anova(fit1)
SSE1 = anova.1[2,2]
MSE1 = SSE1/(dim(CDI)[1] - 2)

y = function(x){
  fit1$coefficients[1] + fit1$coefficients[2]*x
}

sum((CDI$Number_of_active_physicians - y(CDI$Total_population))^2)/(dim(CDI)[
1] - 2)

# MSE 2
anova.2 = anova(fit2)
SSE2 = anova.2[2,2]
MSE2 = SSE2/(dim(CDI)[1] - 2)
```

```

#MSE 3
anova.3 = anova(fit3)
SSE3 = anova.3[2,2]
MSE3 = SSE3/(dim(CDI)[1] - 2)

c(MSE1,MSE2,MSE3)

1.44
a.
CDI[,17] = factor(CDI[,17], levels = c(1, 2, 3, 4),labels = c('NE', 'NC', 'S', 'W'))
data.Geographic_region = split(CDI, CDI$Geographic_region)
level = levels(CDI$Geographic_region)
CDI.NE = data.Geographic_region$NE
CDI.NC = data.Geographic_region$NC
CDI.S = data.Geographic_region$S
CDI.W = data.Geographic_region$W
b.
NE.lm = lm(Per_capita_income ~ Percent_bachelors_degrees, data = CDI.NE)
NE.lm$coefficients

NC.lm = lm(Per_capita_income ~ Percent_bachelors_degrees, data = CDI.NC)
NC.lm$coefficients

S.lm = lm(Per_capita_income ~ Percent_bachelors_degrees, data = CDI.S)
S.lm$coefficients

W.lm = lm(Per_capita_income ~ Percent_bachelors_degrees, data = CDI.W)
W.lm$coefficients

c.
# MSE for region NC
anova(NC.lm)[2,2]/(dim(CDI.NC)[1] - 2)

# MSE for region NE
anova(NE.lm)[2,2]/(dim(CDI.NE)[1] - 2)

# MSE for region S
anova(S.lm)[2,2]/(dim(CDI.S)[1] - 2)

# MSE for region W
anova(W.lm)[2,2]/(dim(CDI.W)[1] - 2)

2.62
# R^2 for Total Population
summary(fit1)$r.squared

```

```

# R^2 for Number of hospital beds
summary(fit2)$r.squared

# R^2 for Total personal income
summary(fit3)$r.squared

2.63
## 90% confidence intervals for each region NC, NE, S, and W, respectively:

confint(NC.lm, 'Percent_bachelors_degrees', level = 0.90)

confint(NE.lm, 'Percent_bachelors_degrees', level = 0.90)

confint(S.lm, 'Percent_bachelors_degrees', level = 0.90)

confint(W.lm, 'Percent_bachelors_degrees', level = 0.90)
anova(NE.lm)
anova(NC.lm)
anova(S.lm)
anova(W.lm)

3.25
par(mar=c(1,1,1,1))
fit.res1 = residuals(fit1)
plot(CDI$Total_population, fit.res1, xlab = 'Total Population', ylab = 'Residuals',
     main = 'Residual Plot for Total Population Model')

qqnorm(fit.res1)
qqline(fit.res1, col = 'red')

fit.res2 = residuals(fit2)
plot(CDI$Number_of_hospital_beds, fit.res2, xlab = 'Number of hospital beds',
     ylab = 'Residuals',
     main = 'Residual Plot for Number of hospital beds Model')

qqnorm(fit.res2)
qqline(fit.res2, col = 'red')

fit.res3 = residuals(fit3)
plot(CDI$Total_personal_income, fit.res3, xlab = 'Total Personal Income', ylab = 'Residuals',
     main = 'Residual Plot for Total Personal Income Model')

qqnorm(fit.res3)
qqline(fit.res3, col = 'red')

```