

Cancer EDA Lina + Duda

Setup

First, we load the car library, which gives us a convenient scatterplotMatrix function.

```
library(car)

## Loading required package: carData
# Load the data
cancer.df=read.csv("cancer.csv")
```

Data Transformation

We're going to explore a set of variables that represent the levels of health insurance coverage for individual counties. There are three variables in the original dataset that are related to insurance:

Variable Name	Description
PctPrivateCoverage	Percentage of the population with private insurance coverage
PctPublicCoverage	Percentage of the population with public insurance coverage
PctEmpPrivCoverage	Percentage of the population with employer-sponsored private insurance coverage

For the purposes of our explanatory analysis, we would like to conduct a more comprehensive research on various types and levels of insurance coverage and their effects on the mortality rates, so it makes sense to define a few more variables that can be derived from the original dataset. For example, we would like to include data about the populations with no insurance coverage, as well as the observations where individuals have both private and public insurance. It can also be more revealing to treat the employer-sponsored coverage as a relative proportion of the private coverage rather than an absolute value.

Hence, let's introduce three new variables as follows:

Variable Name	Description
PctPNoCoverage	Percentage of the population with no insurance coverage
PctDoubleCoverage	Percentage of the population with both private and public insurance coverage
EmpSponsoredPct	Percentage of the private insurance sponsored by employers

We will now add these new variables to our original dataset:

```
nrow(cancer.df[(cancer.df$PctPublicCoverage + cancer.df$PctPrivateCoverage)>100,])

## [1] 1722

nrow(cancer.df[(cancer.df$PctPublicCoverage + cancer.df$PctPrivateCoverage)<100,])

## [1] 1313

cancer.df$PctDoubleCoverage=cancer.df$PctPublicCoverage + cancer.df$PctPrivateCoverage - 100
cancer.df$PctDoubleCoverage[cancer.df$PctDoubleCoverage < 0] = 0
summary(cancer.df$PctDoubleCoverage)

##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
```

```
##    0.000    0.000    1.300    3.203    5.800   31.700
cancer.df$PctNoCoverage = 100 - cancer.df$PctPublicCoverage - cancer.df$PctPrivateCoverage
cancer.df$PctNoCoverage[cancer.df$PctNoCoverage < 0] = 0
summary(cancer.df$PctNoCoverage)

##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##    0.000  0.000   0.000   2.595   3.750   34.600

cancer.df$EmpSponsoredPct = cancer.df$PctEmpPrivCoverage / cancer.df$PctPrivateCoverage * 100
summary(cancer.df$EmpSponsoredPct)

##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##    21.59   59.08   65.14   63.76   69.43   84.55
```

Univariate Analysis of Key Variables

Our key variables in this investigation will be deathRate (target variable) and several independent variables representing insurance coverage for counties' populations.

Cancer Mortality Rate (deathRate variable)

Let's start with the target variable and summarize it:

```
summary(cancer.df$deathRate)

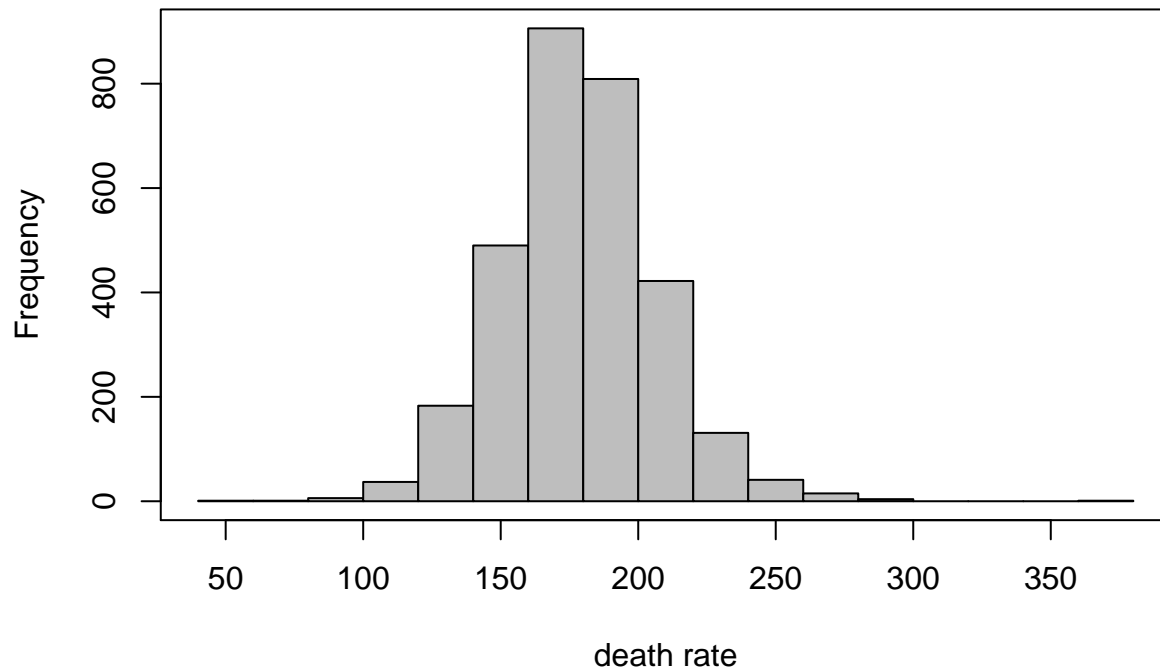
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##    59.7   161.2   178.1   178.7   195.2   362.8
```

We see that this is a metric variable with its mean and median values very close to each other. There are no missing values and no obviously wrong or suspicious outliers.

To better visualize the variable's values distribution, we plot the histogram.

```
with(cancer.df, hist(deathRate, col = "gray",
                     main="Histogram of Cancer Death Rates",
                     xlab="death rate"))
box()
```

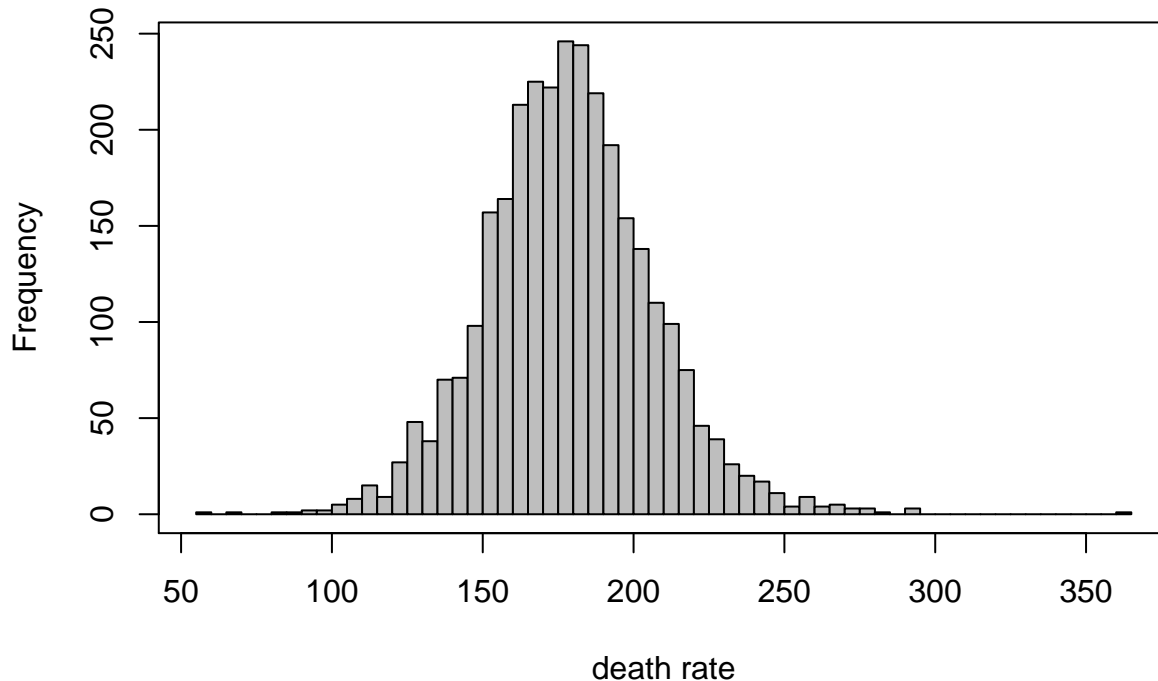
Histogram of Cancer Death Rates



As we can see from the output, the default method for selecting the number of bins produced too few bins, which might obscure some interesting features in the data. A better result is achieved by setting the binning rule to the one proposed by Freedman and Diaconis. Fortunately, `hist()` function has a built-in option for this:

```
with(cancer.df, hist(deathRate, breaks='FD', col = "gray",  
                     main="Histogram of Cancer Death Rates",  
                     xlab="death rate"))  
box()
```

Histogram of Cancer Death Rates



Now

we have a much higher level of detail and can easily infer that deathRate variable distribution is very close to the normal one, with a notable outliers on the far right of the histogram.

Let's explore the extreme outliers with deathRate over 300 and see if we can find anything unusual in these observations. To find out how many outliers are there, we'll use the `nrow()` function:

```
nrow(cancer.df[cancer.df$deathRate > 300,])
```

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

Turns out there's only one observation with this property, so let's examine it a bit closer.

```
str(cancer.df[cancer.df$deathRate > 300,])
```

```
## 'data.frame':    1 obs. of  33 variables:
## $ X                : int 1490
## $ avgAnnCount       : num 214
## $ medIncome         : int 40207
## $ popEst2015        : int 15234
## $ povertyPercent    : num 24.3
## $ binnedInc         : Factor w/ 10 levels "(34218.1, 37413.8]",...: 2
## $ MedianAge         : num 40.3
## $ MedianAgeMale     : num 42.3
## $ MedianAgeFemale   : num 36.9
## $ Geography         : Factor w/ 3047 levels "Abbeville County, South Carolina",...: 2762
## $ AvgHouseholdSize  : num 2.58
## $ PercentMarried    : num 36.4
## $ PctNoHS18_24      : num 27
## $ PctHS18_24        : num 45.1
## $ PctSomeCol18_24   : num NA
## $ PctBachDeg18_24   : num 0
## $ PctHS25_Over      : num 37.4
```

```
## $ PctBachDeg25_Over : num 5.5
## $ PctEmployed16_Over : num NA
## $ PctUnemployed16_Over: num 11.7
## $ PctPrivateCoverage : num 59.6
## $ PctEmpPrivCoverage : num 41
## $ PctPublicCoverage : num 35.8
## $ PctWhite : num 74
## $ PctBlack : num 21.6
## $ PctAsian : num 0.645
## $ PctOtherRace : num 1.53
## $ PctMarriedHouseholds: num 50
## $ BirthRate : num 3.74
## $ deathRate : num 363
## $ PctDoubleCoverage : num 0
## $ PctNoCoverage : num 4.6
## $ EmpSponsoredPct : num 68.8
```

At first sight, nothing in the rest of the data stands out to provide a possible explanation for the high mortality rate (363). We might want to revisit this observation once we completed the rest of the analysis.

Private Insurance Coverage (PctPrivateCoverage variable)

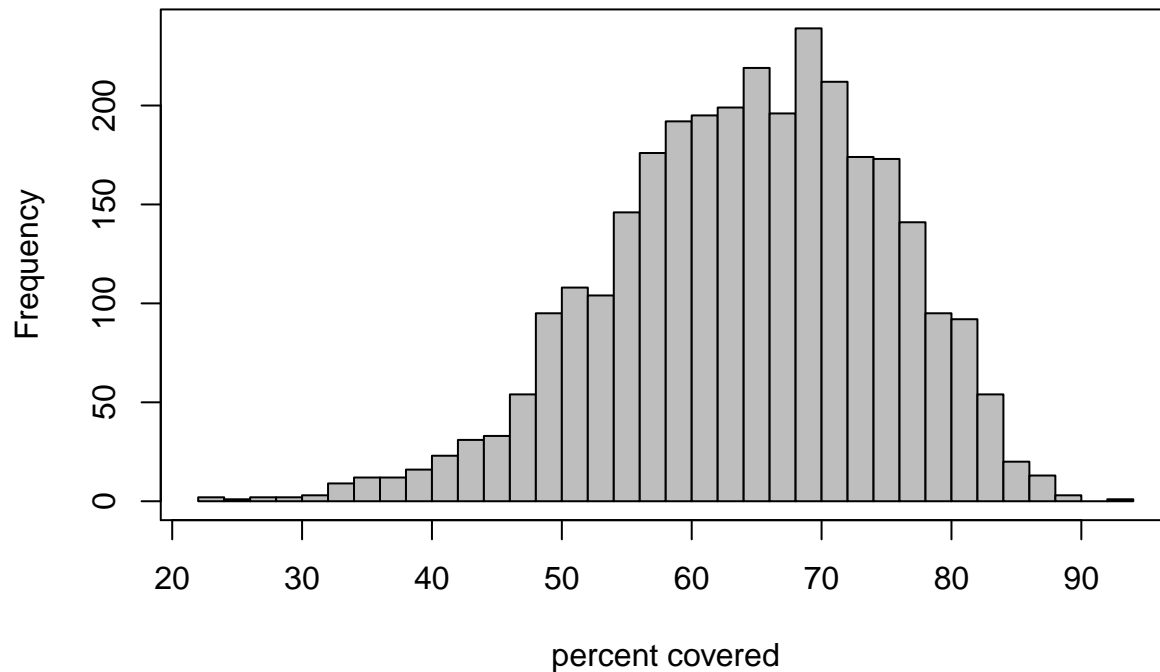
Similar to our target variable, we summarize PctPrivateCoverage and generate its histogram:

```
summary(cancer.df$PctPrivateCoverage)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 22.30   57.20   65.10   64.35   72.10   92.30
```

```
with(cancer.df, hist(PctPrivateCoverage, breaks="FD", col = "gray",
                     main="Histogram of Private Insurance Coverage",
                     xlab="percent covered"))
box()
```

Histogram of Private Insurance Coverage



We notice that the frequency distribution has some negative skew, with the majority of values falling between 55% and 75%. The data looks reasonable, with no obvious errors and missing values.

Public Insurance Coverage (PctPublicCoverage variable)

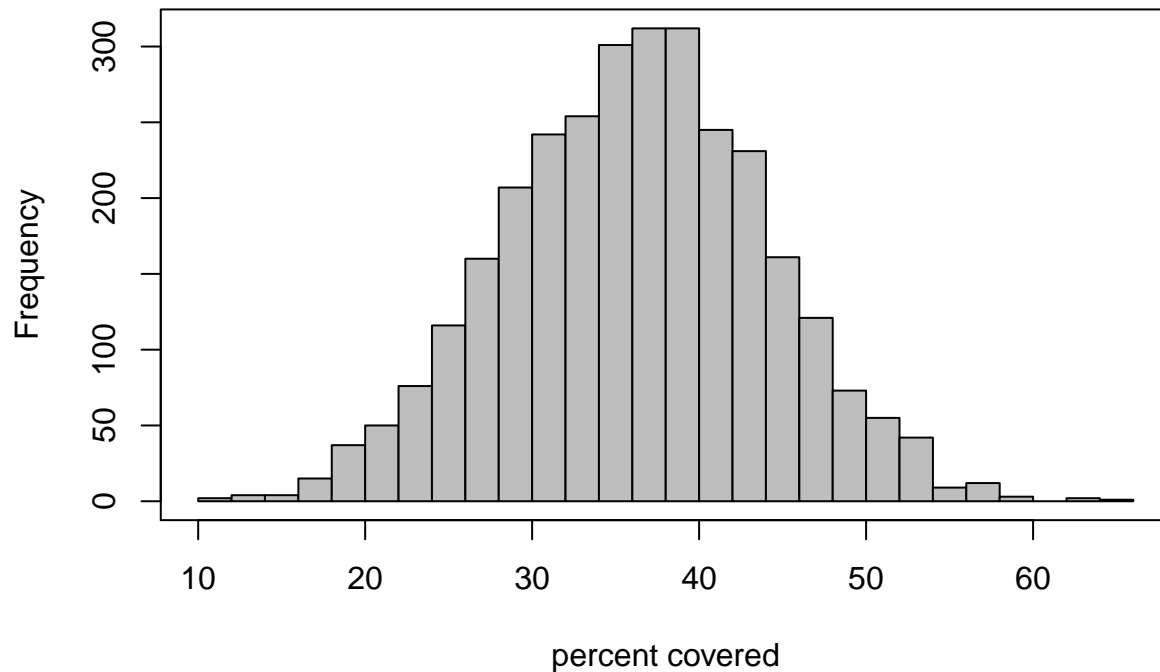
We repeat the steps executed above for the public insurance coverage:

```
summary(cancer.df$PctPublicCoverage)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  11.20   30.90   36.30   36.25  41.55   65.10
```

```
with(cancer.df, hist(PctPublicCoverage, breaks="FD", col = "gray",
  main="Histogram of Public Insurance Coverage",
  xlab = "percent covered"))
box()
```

Histogram of Public Insurance Coverage



Compared to the private insurance coverage, the data is more evenly distributed and is much closer to the normal curve. The mean and median values are almost half of the ones for the private insurance coverage. From that we can infer that the private insurance is much more prevalent than the one sponsored by the state. Similar to PctPrivateCoverage, the public coverage variables doesn't show any obvious errors and there are no missing values.

Employer-sponsored portion of the private coverage (EmpSponsoredPct variable)

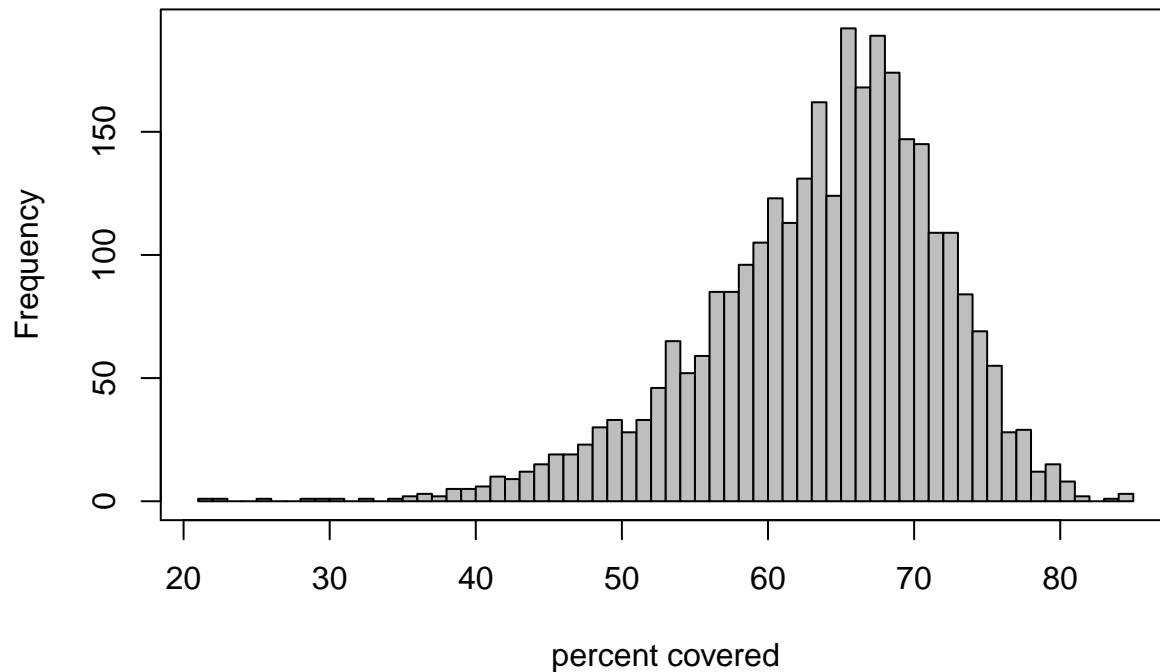
After exploring the general category of the private coverage, we would like to examine what portion of the insurance are provided by employers:

```
summary(cancer.df$EmpSponsoredPct)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  21.59   59.08   65.14   63.76   69.43   84.55
```

```
with(cancer.df, hist(EmpSponsoredPct, breaks="FD", col = "gray",
                      main="Histogram of Employer Portion of Private Coverage",
                      xlab = "percent covered"))
box()
```

Histogram of Employer Portion of Private Coverage



The histogram tells us that employment is the major source of private insurance coverage in the counties: most of the values of EmpSponsoredPct variable fall between 60% and 70%.

No insurance coverage (PctNoCoverage variable)

Let's summarize our generated variable that represents percentage of the population with no insurance coverage:

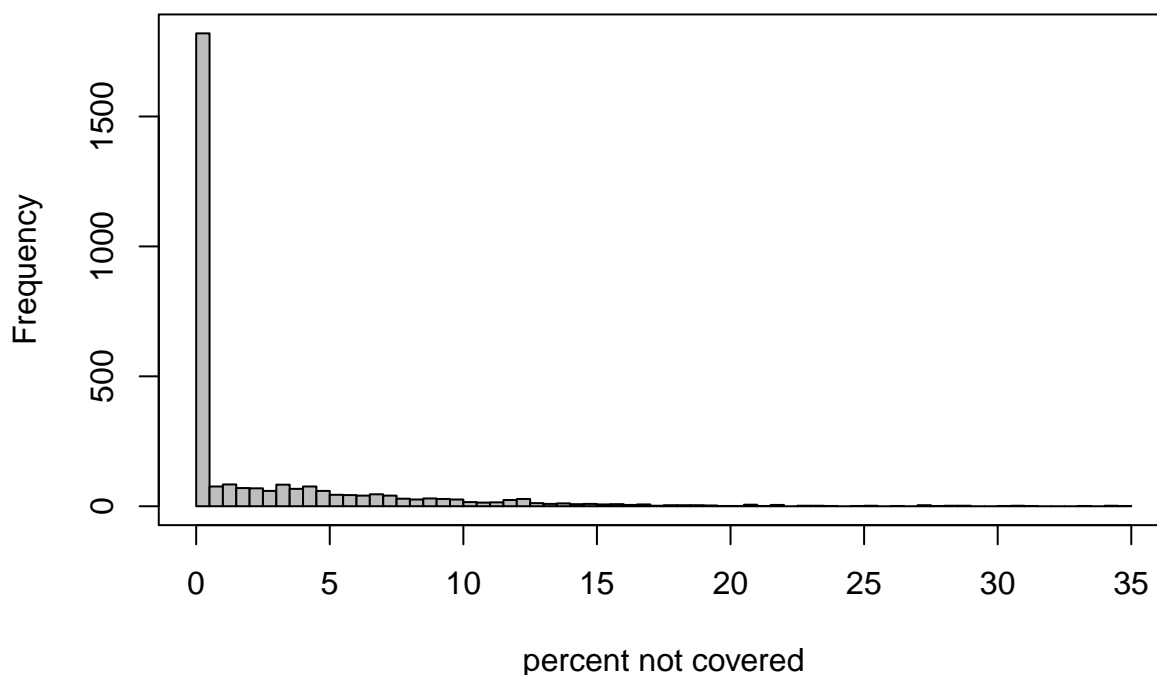
```
summary(cancer.df$PctNoCoverage)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      0.000  0.000   0.000   2.595   3.750   34.600
```

```
with(cancer.df, hist(PctNoCoverage, breaks="FD", col = "gray",
                     main="Histogram of No Insurance Coverage",
                     xlab = "percent not covered"))
```

```
box()
```


Histogram of No Insurance Coverage



Unlike the distributions we've seen so far, this variable has a major peak around 0, with the rest of the values tapering off in the shape of the long-tailed distribution. To get a better insight into the variable, we can generate the percentile metric:

```
quantile(cancer.df$PctNoCoverage, prob = seq(0, 1, length = 11), type = 5)
```

```
##    0%   10%   20%   30%   40%   50%   60%   70%   80%   90%  100%
##  0.0   0.0   0.0   0.0   0.0   0.0   0.6   2.7   4.9   8.7  34.6
```

The result shows that 80% of the observations have less than 5% of the population with no health insurance. We can safely infer then that the effect of this variable on the target will be minimal.

Coverage that includes both private and public components (PctDoubleCoverage variable)

We repeat the steps executed during the evaluation of PctNoCoverage variable:

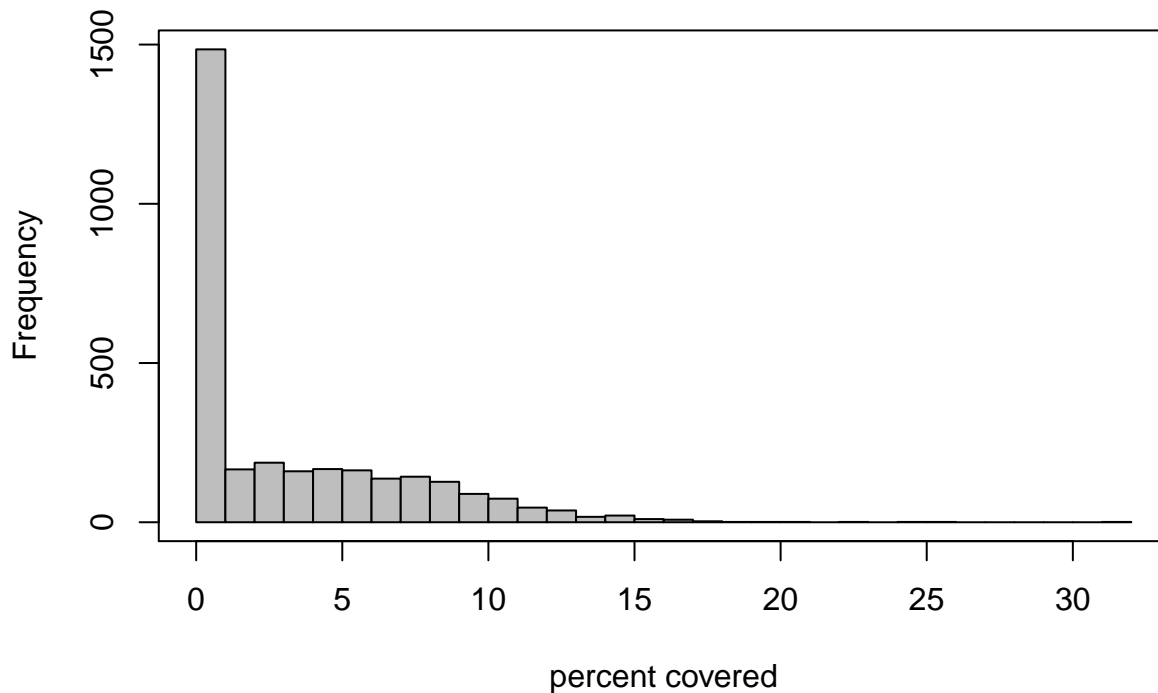
```
summary(cancer.df$PctDoubleCoverage)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  0.000   0.000   1.300   3.203   5.800  31.700
```

```
with(cancer.df, hist(PctDoubleCoverage, breaks="FD", col = "gray",
                      main="Histogram of Double Coverage",
                      xlab = "percent covered"))
```

```
box()
```

Histogram of Double Coverage



```
quantile(cancer.df$PctDoubleCoverage, prob = seq(0, 1, length = 11), type = 5)
```

```
##  0%  10%  20%  30%  40%  50%  60%  70%  80%  90% 100%
##  0.0  0.0  0.0  0.0  0.0  1.3  3.0  4.8  6.9  9.1 31.7
```

The result shows that 80% of the counties have less than 7% of the population with double health insurance. Therefore, similar to the previous case, its relative effect on the target variable will be minimal.

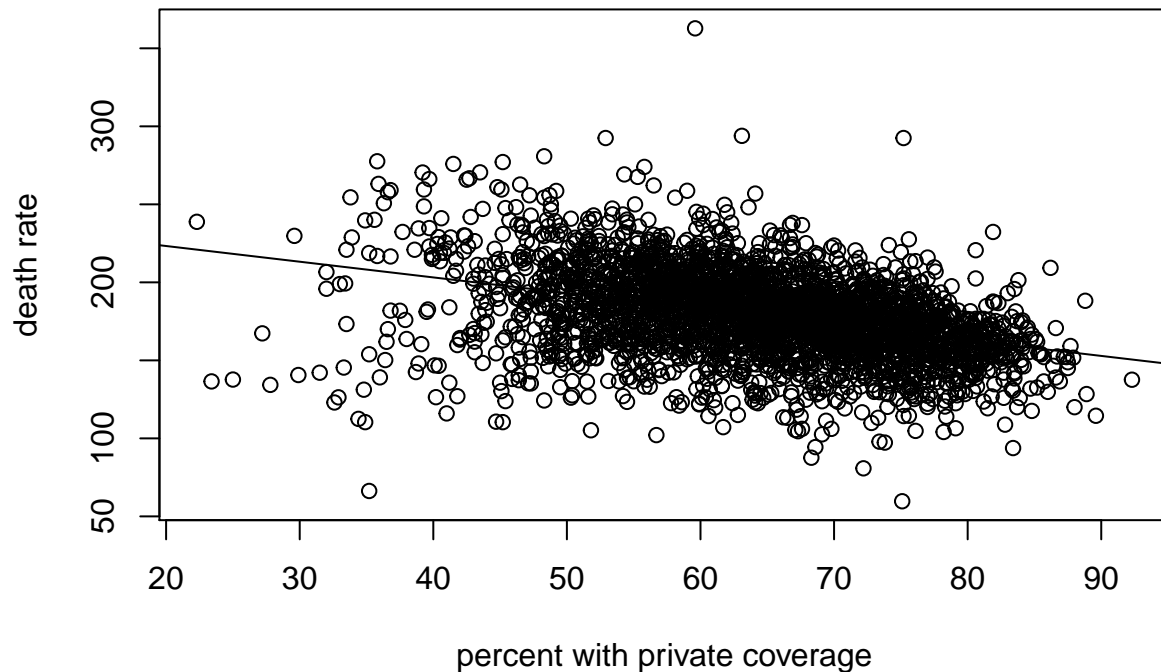
Analysis of Key Relationships

Mortality rates for different levels of private insurance coverage

Our first question is whether having access to a private insurance coverage is correlated with cancer mortality rates. A reasonable hypothesis would be that a cancer patient with a private insurance would be able to afford better treatment options. As a result, she or he will have better chances of survival, so we should expect negative correlation between deathRate and PctPrivateCoverage. Let's build a scatterplot showing the relationship between these two variables. In order to get a better insight into what linear relationship exists in the data, we add the ordinary least squares regression line to the plot and calculate the correlation.

```
plot(cancer.df$PctPrivateCoverage, cancer.df$deathRate,
     xlab = "percent with private coverage", ylab = "death rate",
     main = "Death rates for different levels of private insurance coverage")
abline(lm(cancer.df$deathRate ~ cancer.df$PctPrivateCoverage))
```

Death rates for different levels of private insurance coverage



```
cor(cancer.df$deathRate, cancer.df$PctPrivateCoverage)
```

```
## [1] -0.3860655
```

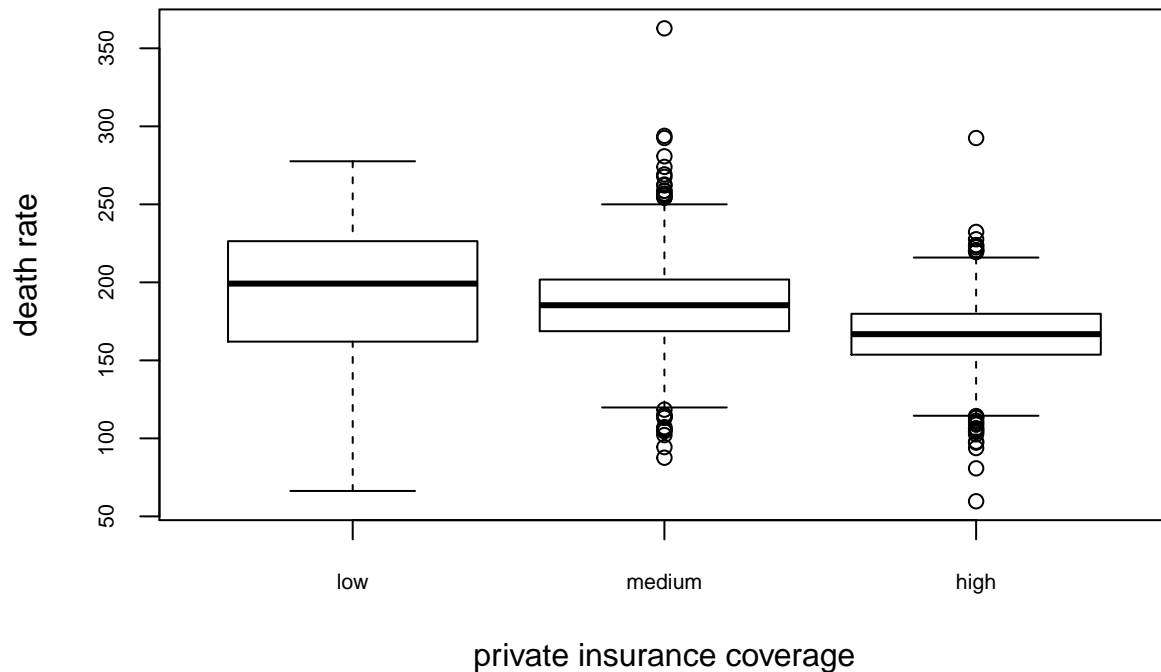
Both from the plot and from the correlation value (-0.39) we can see that they're in agreement with our original hypothesis that mortality rates are lower for the populations with higher percentage of private insurance coverage. The relationship does appear to be linear from about 40% of coverage onward (this is where the majority of observations seem to fall). At the lower end of the graph, the spread of values is much higher. Despite showing the overall trend, the scatterplot is quite noisy, so we might want to confirm our conclusion by generating boxplots for different categories of coverage. First, we'll split the range of PctPrivateCoverage variables into three bins and label them as "low", "medium", and "high" brackets of private insurance coverage. We then will build three separate boxplots for these categories and see how they're distributed relative to deathRate.

```
levels(cut(cancer.df$PctPrivateCoverage, 3, include.lowest=TRUE))
```

```
## [1] "[22.2,45.6]" "(45.6,69]" "(69,92.4]"
```

```
boxplot(deathRate ~ cut(PctPrivateCoverage, 3, include.lowest=TRUE,
  labels=c("low", "medium", "high")),
  data = cancer.df,
  cex.axis = .7,
  main = "Death Rate for different levels of private insurance coverage",
  xlab = "private insurance coverage", ylab = "death rate")
```

Death Rate for different levels of private insurance coverage



The boxplot shows a clear downward trend from the “medium” to “high” category, with the majority of values clustered around the median. The “low” category boxplot, on the other hand, has a much wider spread of data points. We might conclude, therefore, that the effect of private insurance on mortality rates is only noticable for the percentage of coverage which is above certain threshold (~40%). The “medium” category also includes the high death rate outlier we’ve identified earlier (>350). Therefore, the high mortality rate can’t be explained by the inadequate private insurance coverage.

Summary of observations:

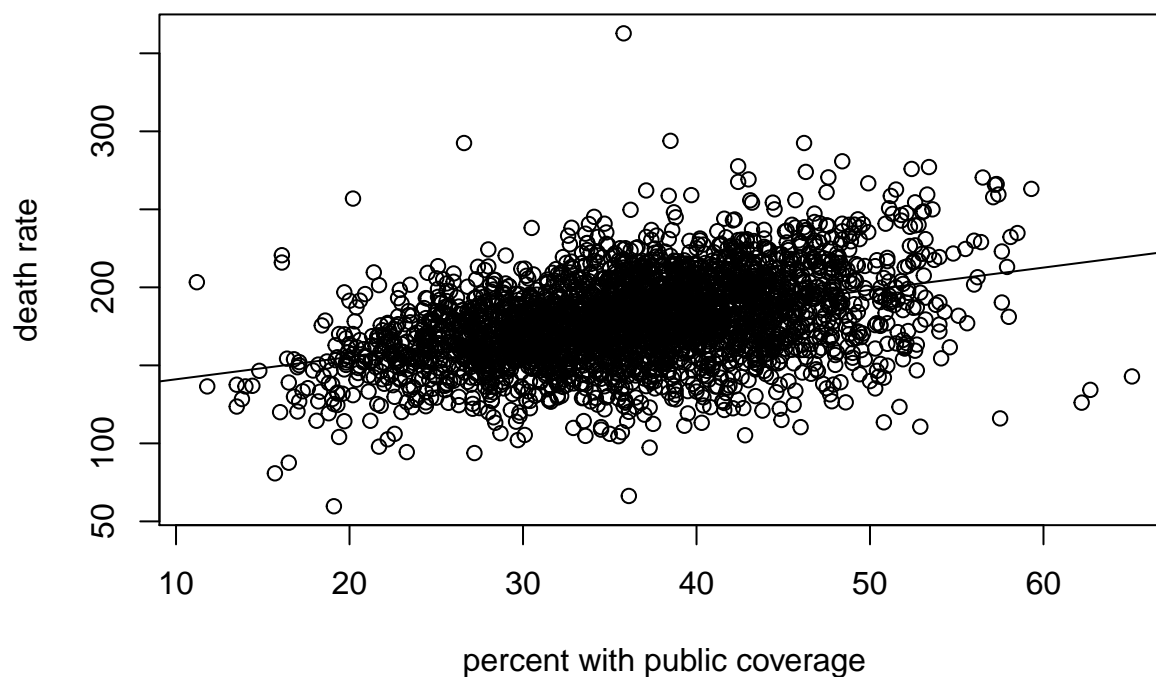
1. There’s a mild negative correlation between cancer mortality rates and access to the private insurance coverage
2. The effect of negative correlation becomes noticable only after the coverage percentage reaches ~40%. Below this point, the data spread is much wider and the effect of private coverage is not obvious.

Mortality rates for different levels of public insurance coverage

We now explore whether public insurance coverage has a similar effect on cancer mortality rates. We repeat the same steps of data analysis we’ve performed for the private insurance variable:

```
plot(cancer.df$PctPublicCoverage, cancer.df$deathRate,  
      xlab = "percent with public coverage", ylab = "death rate",  
      main = "Death rates for different levels of public insurance coverage")  
abline(lm(cancer.df$deathRate ~ cancer.df$PctPublicCoverage))
```

Death rates for different levels of public insurance coverage



```
cor(cancer.df$deathRate, cancer.df$PctPublicCoverage)
```

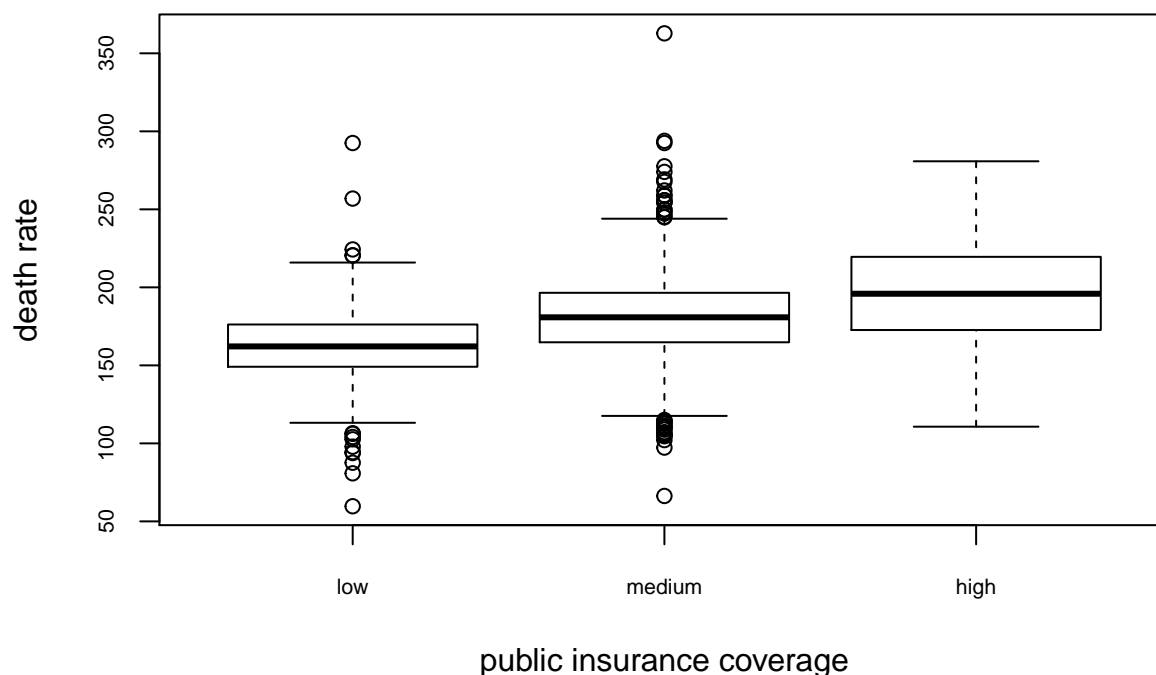
```
## [1] 0.4045717
```

```
levels(cut(cancer.df$PctPublicCoverage, 3, include.lowest=TRUE))
```

```
## [1] "[11.1,29.2]" "(29.2,47.1]" "(47.1,65.2]"
```

```
boxplot(deathRate ~ cut(PctPublicCoverage, 3, include.lowest=TRUE,  
  labels=c("low", "medium", "high")),  
  data = cancer.df,  
  cex.axis = .7,  
  main = "Death Rate for different levels of public insurance coverage",  
  xlab = "public insurance coverage", ylab = "death rate")
```

Death Rate for different levels of public insurance coverage



Contrary to our expectations, we see the directly opposite relationship between public insurance coverage and cancer mortality rates. The values are positively correlated and the correlation's absolute value is even higher than the one we calculated for private insurance coverage. There's also no salient threshold effect we observed earlier: the relationship appears to be linear throughout the entire range of coverage percentage.

Summary of observations:

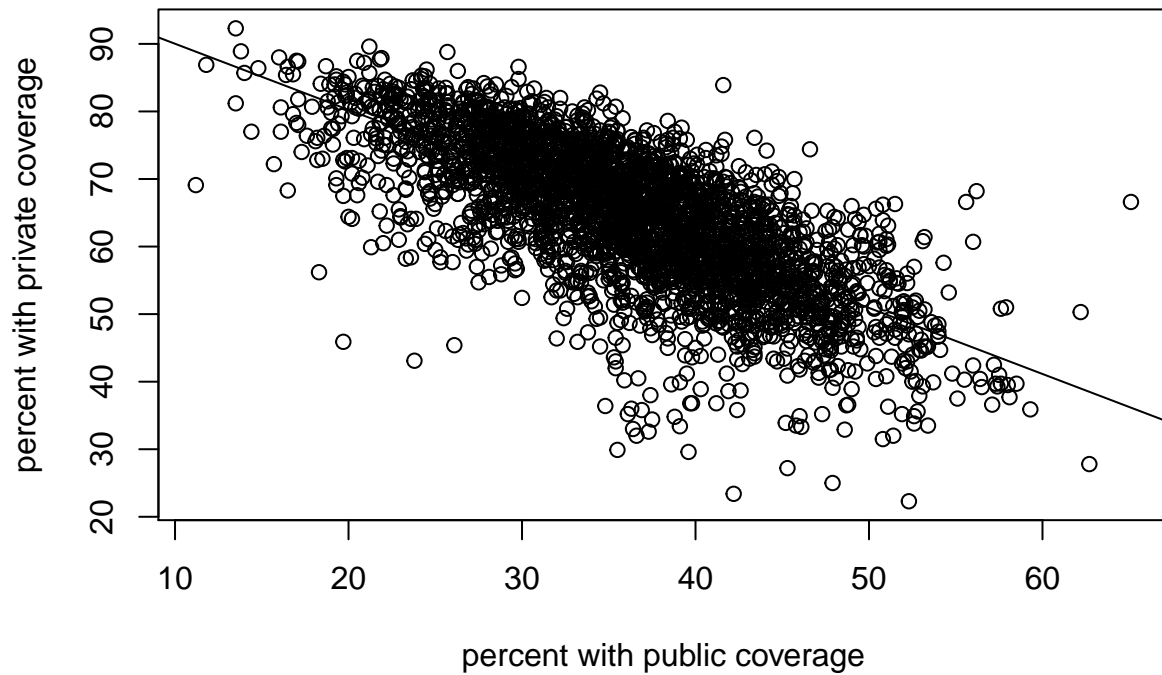
1. There's a noticeable positive correlation between cancer mortality rates and availability of public insurance coverage
2. The relationship is very close to the linear one throughout the entire range of coverage's percentages

Relationship between private and public insurance coverage

We will now explore if there is any meaningful relationship between private and public insurance coverage. As in the earlier steps of our investigation, we generate a scatterplot and box plots for these variables, and compute the correlation value:

```
plot(cancer.df$PctPublicCoverage, cancer.df$PctPrivateCoverage,  
      xlab = "percent with public coverage", ylab = "percent with private coverage",  
      main = "Private coverage for different levels of public insurance coverage")  
abline(lm(cancer.df$PctPrivateCoverage ~ cancer.df$PctPublicCoverage))
```

Private coverage for different levels of public insurance coverage



```
cor(cancer.df$PctPrivateCoverage, cancer.df$PctPublicCoverage)
```

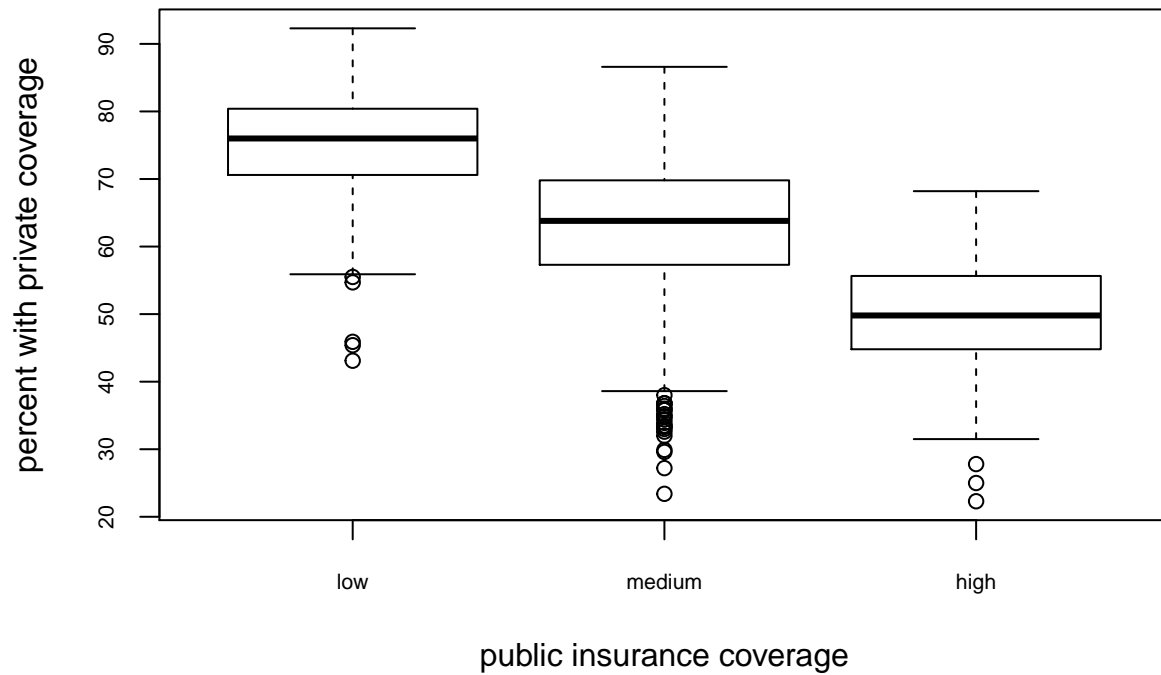
```
## [1] -0.7200115
```

```
levels(cut(cancer.df$PctPublicCoverage, 3, include.lowest=TRUE))
```

```
## [1] "[11.1,29.2]" "(29.2,47.1]" "(47.1,65.2]"
```

```
boxplot(PctPrivateCoverage ~ cut(PctPublicCoverage, 3, include.lowest=TRUE,  
  labels=c("low", "medium", "high")),  
  data = cancer.df,  
  cex.axis = .7,  
  main = "Private coverage for different levels of public insurance coverage",  
  xlab = "public insurance coverage", ylab = "percent with private coverage")
```

Private coverage for different levels of public insurance coverage



Summary of observations:

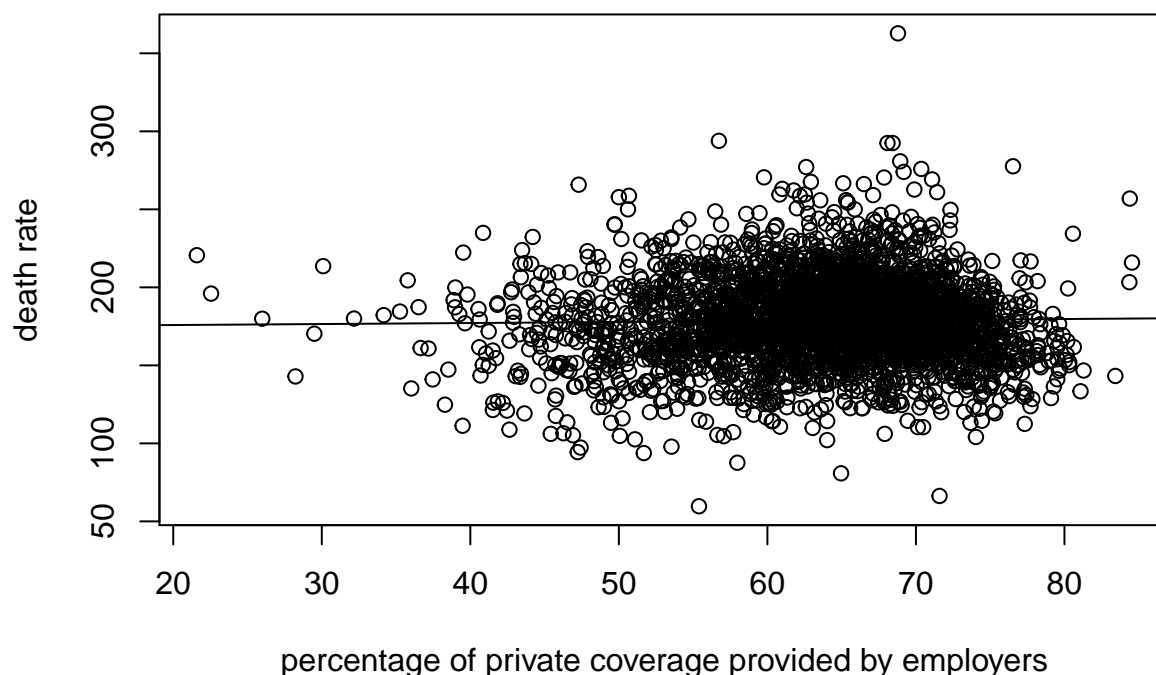
1. There's a strong negative correlation between private and public insurance coverage
2. The majority of observations cluster around ordinary least squares regression line, emphasizing linear relationship between the two variables

Mortality rates for different levels of employer-sponsored private coverage

Finally, let's see if the relative portion of employer-sponsored private insurance coverage has any relationship with cancer mortality rates.

```
plot(cancer.df$EmpSponsoredPct, cancer.df$deathRate,  
      xlab = "percentage of private coverage provided by employers",  
      ylab = "death rate",  
      main = "Death rates for different levels of employer coverage")  
abline(lm(cancer.df$deathRate ~ cancer.df$EmpSponsoredPct))
```


Death rates for different levels of employer coverage



```
cor(cancer.df$deathRate, cancer.df$EmpSponsoredPct)
```

```
## [1] 0.01885173
```

Summary of observations:

1. From the data analysis above, we don't detect any noticeable relationships between the cancer mortality rates and the composition of the private insurance coverage.

Conclusion: insurance coverage per ce doesn't improve cancer mortality rates

1. The data analysis we performed has refuted our hypothesis that cancer patients who have access to health insurance have better chances of survival.
2. We also saw that private and public insurance demonstrate opposite relationships with cancer mortality rates.
3. One of the important findings is that higher levels of public insurance coverage are strongly correlated with lower percentage of private insurance coverage.
4. The relative amount of private coverage sponsored by employers have no detectable relationships with cancer mortality rates.
5. In order to explain these counter-intuitive results, as well as the 'threshold effect' of private coverage, we need to explore other variables that might directly influence these relationships.

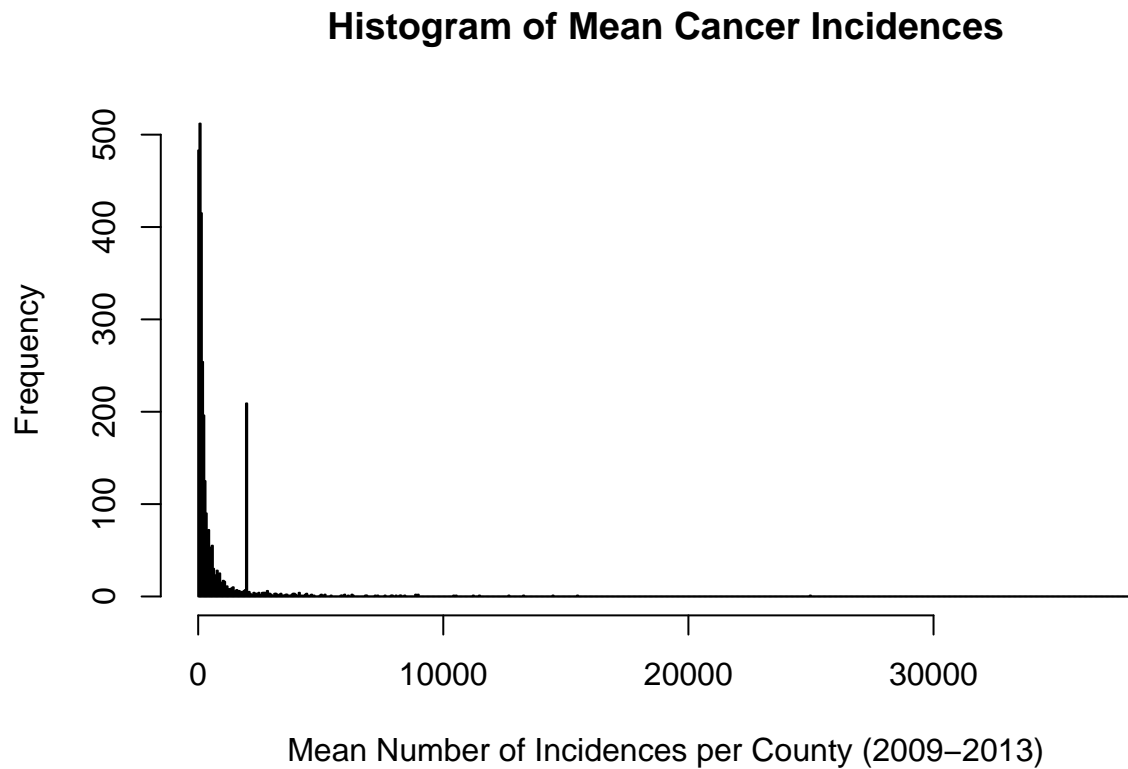
Analysis of Secondary Effects

Since we have seen that private and public insurance have opposite relationships with cancer mortality rates and highly negatively correlated between each other, we now must explore other variables that could be influencing these relationships.

We will start with a univariate analysis of other selected variables

avgAnnCount

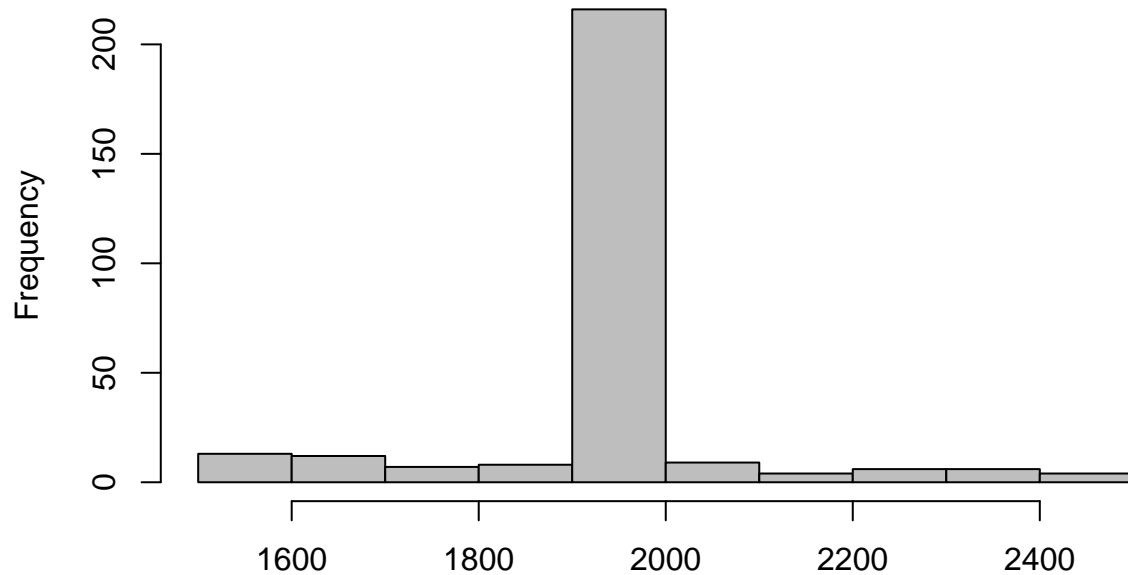
```
hist(cancer.df$avgAnnCount, breaks="fd", xlab = "Mean Number of Incidences per County (2009-2013)", ylab = "Frequency")
```



That second peak seems a little odd. We should take a look at what is going on there, where the *avgAnnCount* is between 1500 and 2500

```
hist(cancer.df$avgAnnCount[cancer.df$avgAnnCount>1500 & cancer.df$avgAnnCount<2500], xlab="avgAnnCount between 1500 and 2500", ylab="Frequency")
```

Histogram of avgAnnCount outliers



avgAnnCount between 1500 and 2500

We can now see that the peaks are between avgAnnCount of 1900 and 2000, a range easier to investigate:

```
cancer.df[cancer.df$avgAnnCount>1900 & cancer.df$avgAnnCount<2000,]
```

##	X	avgAnnCount	medIncome	popEst2015	povertyPercent
## 112	112	1962.668	46840	24200	13.2
## 113	113	1962.668	51241	2114801	15.6
## 114	114	1962.668	60322	47710	10.3
## 115	115	1962.668	72648	51935	10.8
## 116	116	1962.668	47180	829	14.0
## 117	117	1962.668	70535	2016	9.8
## 118	118	1962.668	67423	17019	10.7
## 119	119	1962.668	74347	5903	11.0
## 120	120	1962.668	45629	5036	15.1
## 121	121	1962.668	48576	52585	13.7
## 122	122	1962.668	40714	4478	19.0
## 123	123	1962.668	42881	42477	16.8
## 124	124	1962.668	52001	6634	18.5
## 125	125	1962.668	55291	3987	8.4
## 126	126	1962.668	52862	446903	15.4
## 127	127	1962.668	57243	9811	13.7
## 128	128	1962.668	45639	54521	19.3
## 149	149	1997.000	73750	371398	11.8
## 368	368	1962.668	40676	15702	14.9
## 369	369	1962.668	70868	344151	7.6
## 370	370	1962.668	51116	33386	14.6
## 371	371	1962.668	43706	45672	18.5
## 372	372	1962.668	52531	39710	13.6
## 568	568	1908.000	45698	392664	16.1
## 1014	1014	1948.000	75689	375391	8.3

##	1026	1026	1912.000	62116	444769	17.4
##	1295	1295	1962.668	50886	2640	10.4
##	1296	1296	1962.668	45341	2591	12.0
##	1297	1297	1962.668	60279	7733	10.9
##	1298	1298	1962.668	58543	6133	8.3
##	1299	1299	1962.668	52795	1330	10.8
##	1300	1300	1962.668	39732	6244	16.9
##	1301	1301	1962.668	47709	2474	12.8
##	1302	1302	1962.668	50525	5817	12.9
##	1303	1303	1962.668	55074	35073	12.8
##	1304	1304	1962.668	59596	4064	10.7
##	1305	1305	1962.668	52394	1893	8.8
##	1306	1306	1962.668	53796	13338	10.3
##	1307	1307	1962.668	56927	18930	9.4
##	1308	1308	1962.668	44101	2970	13.5
##	1309	1309	1962.668	76104	580159	6.6
##	1310	1310	1962.668	53484	3956	12.2
##	1311	1311	1962.668	53378	7687	11.2
##	1312	1312	1962.668	46318	2564	12.8
##	1313	1313	1962.668	53739	1670	10.9
##	1314	1314	1962.668	65549	79315	10.9
##	1315	1315	1962.668	44896	3105	12.7
##	1316	1316	1962.668	44520	9536	14.8
##	1317	1317	1962.668	49852	2825	9.7
##	1318	1318	1962.668	43038	33339	18.0
##	1319	1319	1962.668	57878	28941	8.8
##	1320	1320	1962.668	47723	12103	11.9
##	1321	1321	1962.668	47901	9936	11.1
##	1322	1322	1962.668	53903	4330	11.7
##	1323	1323	1962.668	63924	32553	9.5
##	1324	1324	1962.668	48645	6282	11.2
##	1325	1325	1962.668	42221	33314	18.0
##	1326	1326	1962.668	44774	5645	12.5
##	1327	1327	1962.668	55378	3007	11.1
##	1328	1328	1962.668	54149	10227	9.9
##	1329	1329	1962.668	41527	16346	17.7
##	1330	1330	1962.668	51923	3005	10.0
##	1331	1331	1962.668	44846	5550	13.5
##	1332	1332	1962.668	52749	15847	12.9
##	1333	1333	1962.668	38848	3683	12.9
##	1334	1334	1962.668	52744	5975	10.4
##	1335	1335	1962.668	47351	6838	13.9
##	1336	1336	1962.668	45856	5428	11.4
##	1337	1337	1962.668	65373	23298	10.0
##	1338	1338	1962.668	54701	9691	10.3
##	1339	1339	1962.668	44932	2506	11.8
##	1340	1340	1962.668	46533	63718	15.8
##	1341	1341	1962.668	42381	4725	12.7
##	1342	1342	1962.668	49152	9977	14.7
##	1343	1343	1962.668	46468	75247	20.2
##	1344	1344	1962.668	48383	5174	12.0
##	1345	1345	1962.668	43083	3130	12.1
##	1346	1346	1962.668	45443	7039	13.3
##	1347	1347	1962.668	48460	55691	13.0

##	1348	1348	1962.668	56032	4964	8.3
##	1349	1349	1962.668	51175	511574	14.7
##	1350	1350	1962.668	47381	23152	15.7
##	1351	1351	1962.668	52795	178725	15.0
##	1352	1352	1962.668	52300	2512	10.9
##	1353	1353	1962.668	43100	5983	16.2
##	1354	1354	1962.668	41303	3704	13.5
##	1355	1355	1962.668	45945	4236	14.1
##	1356	1356	1962.668	56591	2072	11.5
##	1357	1357	1962.668	59773	5806	9.7
##	1358	1358	1962.668	52476	23535	13.3
##	1359	1359	1962.668	50639	7904	11.3
##	1360	1360	1962.668	46859	2927	9.9
##	1361	1361	1962.668	59160	6951	9.6
##	1362	1362	1962.668	49109	1518	12.0
##	1363	1363	1962.668	43264	5598	10.8
##	1364	1364	1962.668	53447	2157	11.5
##	1365	1365	1962.668	39764	8856	16.7
##	1366	1366	1962.668	35502	3115	18.4
##	1367	1367	1962.668	37087	163369	24.4
##	1486	1486	1940.000	56443	449144	11.8
##	1564	1564	1976.000	65555	401515	9.5
##	1744	1744	1931.000	61510	300813	10.0
##	1840	1840	1931.000	88516	716087	8.8
##	1924	1924	1929.000	70929	537559	10.9
##	2454	2454	1962.668	46495	5040	12.8
##	2455	2455	1962.668	49513	65787	17.9
##	2456	2456	1962.668	52272	25313	8.9
##	2457	2457	1962.668	53291	35569	12.1
##	2458	2458	1962.668	88500	98741	4.8
##	2459	2459	1962.668	45179	28706	15.7
##	2460	2460	1962.668	51274	12109	11.5
##	2461	2461	1962.668	75122	54293	6.2
##	2462	2462	1962.668	55560	62324	13.3
##	2463	2463	1962.668	43674	8803	17.8
##	2464	2464	1962.668	48593	5194	10.4
##	2465	2465	1962.668	48176	11549	12.9
##	2466	2466	1962.668	50162	63428	10.9
##	2467	2467	1962.668	76269	414686	7.3
##	2468	2468	1962.668	68777	20364	7.6
##	2469	2469	1962.668	52298	37075	9.6
##	2470	2470	1962.668	48163	14050	13.5
##	2471	2471	1962.668	52444	20834	11.6
##	2472	2472	1962.668	47203	30613	12.5
##	2473	2473	1962.668	60869	46435	9.6
##	2474	2474	1962.668	48179	5903	10.5
##	2475	2475	1962.668	64490	1223149	13.0
##	2476	2476	1962.668	59900	18773	8.4
##	2477	2477	1962.668	47944	20655	12.2
##	2478	2478	1962.668	65342	38429	7.8
##	2479	2479	1962.668	48525	45435	14.7
##	2480	2480	1962.668	53700	10079	9.8
##	2481	2481	1962.668	48908	15837	13.8
##	2482	2482	1962.668	57405	42542	11.5

##	2483	2483	1962.668	50372	4424	9.9
##	2484	2484	1962.668	44113	12841	15.3
##	2485	2485	1962.668	48245	6856	10.4
##	2486	2486	1962.668	52381	10631	10.5
##	2487	2487	1962.668	44727	3923	10.7
##	2488	2488	1962.668	61665	27663	8.9
##	2489	2489	1962.668	48449	5771	10.2
##	2490	2490	1962.668	53552	25673	11.5
##	2491	2491	1962.668	60114	35932	7.9
##	2492	2492	1962.668	39926	5457	19.6
##	2493	2493	1962.668	54433	9423	9.6
##	2494	2494	1962.668	52042	20022	12.2
##	2495	2495	1962.668	54354	23102	10.0
##	2496	2496	1962.668	48434	25788	12.0
##	2497	2497	1962.668	50685	32775	11.8
##	2498	2498	1962.668	47537	39116	12.1
##	2499	2499	1962.668	52219	8413	9.6
##	2500	2500	1962.668	61279	33347	10.2
##	2501	2501	1962.668	50684	21770	13.0
##	2502	2502	1962.668	51597	6678	13.4
##	2503	2503	1962.668	69430	151436	9.8
##	2504	2504	1962.668	51510	57716	11.2
##	2505	2505	1962.668	50905	14219	9.3
##	2506	2506	1962.668	42157	29069	17.5
##	2507	2507	1962.668	47164	9271	12.2
##	2508	2508	1962.668	50175	31533	12.7
##	2509	2509	1962.668	51343	11041	10.4
##	2510	2510	1962.668	55070	538133	16.5
##	2511	2511	1962.668	48188	4055	9.2
##	2512	2512	1962.668	49934	15471	12.4
##	2513	2513	1962.668	54386	14892	11.3
##	2514	2514	1962.668	60317	65400	13.0
##	2515	2515	1962.668	55352	9600	9.6
##	2516	2516	1962.668	53838	15770	9.2
##	2517	2517	1962.668	49714	200431	17.0
##	2518	2518	1962.668	91688	141660	5.5
##	2519	2519	1962.668	76512	91705	7.2
##	2520	2520	1962.668	59893	14875	9.2
##	2521	2521	1962.668	55832	154708	13.4
##	2522	2522	1962.668	57850	36755	10.9
##	2523	2523	1962.668	55292	9796	14.2
##	2524	2524	1962.668	51032	9340	10.8
##	2525	2525	1962.668	43287	24257	17.8
##	2526	2526	1962.668	49536	3401	12.2
##	2527	2527	1962.668	58752	21239	7.3
##	2528	2528	1962.668	41909	13875	15.8
##	2529	2529	1962.668	54183	18989	10.1
##	2530	2530	1962.668	84113	251597	6.0
##	2531	2531	1962.668	48658	10952	10.4
##	2532	2532	1962.668	54536	6396	10.3
##	2533	2533	1962.668	49365	50885	15.8
##	2534	2534	1962.668	76489	131311	6.0
##	2535	2535	1962.668	51028	9875	11.2
##	3019	3019	1962.668	39196	12717	18.2

##	3020	3020	1962.668	45368	7808	14.3
##	3021	3021	1962.668	44199	16398	17.4
##	3022	3022	1962.668	51676	4823	12.4
##	3023	3023	1962.668	46942	27103	17.8
##	3024	3024	1962.668	38430	14712	18.4
##	3025	3025	1962.668	45781	9776	17.3
##	3026	3026	1962.668	60123	66741	11.3
##	3027	3027	1962.668	47025	2679	13.2
##	3028	3028	1962.668	39120	3402	16.4
##	3029	3029	1962.668	41797	2679	11.5
##	3030	3030	1962.668	47650	2096	13.0
##	3031	3031	1962.668	53131	8347	10.9
##	3032	3032	1962.668	42697	9219	14.0
##	3033	3033	1962.668	55705	8384	10.9
##	3034	3034	1962.668	45353	1843	11.4
##	3035	3035	1962.668	45180	35788	15.0
##	3036	3036	1962.668	41434	2932	13.9
##	3037	3037	1962.668	47493	19303	11.1
##	3038	3038	1962.668	47599	7797	15.0
##	3039	3039	1962.668	49246	118053	19.4
##	3040	3040	1962.668	49256	2968	10.5
##	3041	3041	1962.668	36471	2605	17.4
##	3042	3042	1962.668	49508	29029	13.0
##	3043	3043	1962.668	46961	6343	12.4
##	3044	3044	1962.668	48609	37118	18.8
##	3045	3045	1962.668	51144	34536	15.0
##	3046	3046	1962.668	50745	25609	13.3
##	3047	3047	1962.668	41193	37030	13.9
##		binnedInc	MedianAge	MedianAgeMale	MedianAgeFemale	
##	112	(45201, 48021.6]	38.9	37.1	40.8	
##	113	(51046.4, 54545.6]	36.4	35.9	37.0	
##	114	(54545.6, 61494.5]	49.4	48.7	50.0	
##	115	(61494.5, 125635]	33.5	32.9	34.1	
##	116	(45201, 48021.6]	46.5	43.0	47.5	
##	117	(61494.5, 125635]	43.9	44.9	42.6	
##	118	(61494.5, 125635]	35.3	34.9	35.4	
##	119	(61494.5, 125635]	37.2	36.4	37.6	
##	120	(45201, 48021.6]	37.4	37.3	38.1	
##	121	(48021.6, 51046.4]	43.2	43.1	43.4	
##	122	(40362.7, 42724.4]	48.7	48.0	49.2	
##	123	(42724.4, 45201]	50.8	50.8	50.8	
##	124	(51046.4, 54545.6]	41.3	39.5	44.1	
##	125	(54545.6, 61494.5]	54.7	54.0	55.6	
##	126	(51046.4, 54545.6]	37.6	36.8	38.4	
##	127	(54545.6, 61494.5]	39.4	38.4	40.3	
##	128	(45201, 48021.6]	42.6	41.2	44.4	
##	149	(61494.5, 125635]	38.4	36.9	39.9	
##	368	(40362.7, 42724.4]	53.1	52.7	53.4	
##	369	(61494.5, 125635]	37.9	37.2	38.7	
##	370	(51046.4, 54545.6]	42.4	42.0	42.9	
##	371	(42724.4, 45201]	33.3	31.8	34.7	
##	372	(51046.4, 54545.6]	35.4	35.3	35.6	
##	568	(45201, 48021.6]	36.4	34.8	37.9	
##	1014	(61494.5, 125635]	41.0	39.7	42.2	

## 1026	(61494.5, 125635]	33.6	32.4	35.0
## 1295	(48021.6, 51046.4]	47.4	45.3	50.1
## 1296	(45201, 48021.6]	49.8	49.0	51.9
## 1297	(54545.6, 61494.5]	32.8	32.4	33.2
## 1298	(54545.6, 61494.5]	36.1	36.8	35.0
## 1299	(51046.4, 54545.6]	49.4	48.7	49.9
## 1300	(37413.8, 40362.7]	48.0	48.0	48.0
## 1301	(45201, 48021.6]	38.0	35.8	39.1
## 1302	(48021.6, 51046.4]	43.0	39.7	46.7
## 1303	(54545.6, 61494.5]	38.6	37.7	39.3
## 1304	(54545.6, 61494.5]	36.1	32.6	38.5
## 1305	(51046.4, 54545.6]	47.8	46.3	49.3
## 1306	(51046.4, 54545.6]	40.8	39.4	42.5
## 1307	(54545.6, 61494.5]	43.5	42.6	44.6
## 1308	(42724.4, 45201]	52.1	50.7	52.8
## 1309	(61494.5, 125635]	36.8	35.6	38.1
## 1310	(51046.4, 54545.6]	34.8	30.6	39.3
## 1311	(51046.4, 54545.6]	45.2	41.7	46.8
## 1312	(45201, 48021.6]	41.7	41.3	42.4
## 1313	(51046.4, 54545.6]	535.2	44.7	44.4
## 1314	(61494.5, 125635]	37.3	36.4	38.6
## 1315	(42724.4, 45201]	46.2	43.8	47.9
## 1316	(42724.4, 45201]	43.7	44.3	42.9
## 1317	(48021.6, 51046.4]	42.2	40.6	45.7
## 1318	(42724.4, 45201]	32.5	32.5	32.4
## 1319	(54545.6, 61494.5]	42.0	40.2	43.6
## 1320	(45201, 48021.6]	45.0	43.7	46.0
## 1321	(45201, 48021.6]	45.0	43.3	46.3
## 1322	(51046.4, 54545.6]	39.8	36.5	42.4
## 1323	(61494.5, 125635]	40.4	39.5	41.4
## 1324	(48021.6, 51046.4]	44.4	40.8	48.0
## 1325	(40362.7, 42724.4]	39.4	37.0	41.9
## 1326	(42724.4, 45201]	47.6	46.6	49.5
## 1327	(54545.6, 61494.5]	41.6	38.3	46.3
## 1328	(51046.4, 54545.6]	41.6	39.7	43.6
## 1329	(40362.7, 42724.4]	40.0	39.7	40.6
## 1330	(51046.4, 54545.6]	49.2	46.0	51.9
## 1331	(42724.4, 45201]	44.0	42.1	48.5
## 1332	(51046.4, 54545.6]	42.6	42.2	42.9
## 1333	(37413.8, 40362.7]	48.5	46.5	49.7
## 1334	(51046.4, 54545.6]	42.2	39.5	45.4
## 1335	(45201, 48021.6]	43.1	41.5	48.2
## 1336	(45201, 48021.6]	44.7	44.2	45.5
## 1337	(61494.5, 125635]	34.7	34.7	34.6
## 1338	(54545.6, 61494.5]	39.1	35.0	42.6
## 1339	(42724.4, 45201]	50.9	48.7	52.4
## 1340	(45201, 48021.6]	39.5	37.6	42.3
## 1341	(40362.7, 42724.4]	50.9	50.0	51.7
## 1342	(48021.6, 51046.4]	37.9	35.3	41.9
## 1343	(45201, 48021.6]	24.7	24.5	25.3
## 1344	(48021.6, 51046.4]	43.9	43.2	45.3
## 1345	(42724.4, 45201]	48.8	48.1	49.8
## 1346	(45201, 48021.6]	47.1	43.8	48.5
## 1347	(48021.6, 51046.4]	37.7	36.8	39.1

## 1348 (54545.6, 61494.5]	42.0	40.7	43.4
## 1349 (51046.4, 54545.6]	34.6	33.4	35.7
## 1350 (45201, 48021.6]	29.4	29.2	29.6
## 1351 (51046.4, 54545.6]	38.8	37.4	40.1
## 1352 (51046.4, 54545.6]	45.7	43.1	48.3
## 1353 (42724.4, 45201]	38.4	39.4	36.8
## 1354 (40362.7, 42724.4]	50.7	50.3	50.9
## 1355 (45201, 48021.6]	43.7	42.3	45.2
## 1356 (54545.6, 61494.5]	35.7	36.9	33.6
## 1357 (54545.6, 61494.5]	34.9	34.3	37.0
## 1358 (51046.4, 54545.6]	40.8	39.4	41.7
## 1359 (48021.6, 51046.4]	35.3	34.2	37.2
## 1360 (45201, 48021.6]	49.1	48.8	49.4
## 1361 (54545.6, 61494.5]	42.1	41.9	43.0
## 1362 (48021.6, 51046.4]	41.8	43.4	38.8
## 1363 (42724.4, 45201]	46.3	44.1	48.4
## 1364 (51046.4, 54545.6]	35.7	34.5	40.5
## 1365 (37413.8, 40362.7]	43.8	42.1	46.1
## 1366 (34218.1, 37413.8]	47.4	46.9	48.6
## 1367 (34218.1, 37413.8]	33.3	32.4	34.2
## 1486 (54545.6, 61494.5]	39.1	37.1	40.5
## 1564 (61494.5, 125635]	41.3	40.1	42.5
## 1744 (61494.5, 125635]	41.8	40.0	43.2
## 1840 (61494.5, 125635]	35.5	34.6	36.3
## 1924 (61494.5, 125635]	36.4	35.5	37.3
## 2454 (45201, 48021.6]	49.2	47.6	50.4
## 2455 (48021.6, 51046.4]	30.1	29.3	31.0
## 2456 (51046.4, 54545.6]	43.5	42.0	45.2
## 2457 (51046.4, 54545.6]	41.0	40.6	41.8
## 2458 (61494.5, 125635]	36.9	36.1	37.6
## 2459 (42724.4, 45201]	48.3	48.0	48.6
## 2460 (51046.4, 54545.6]	43.0	42.5	44.3
## 2461 (61494.5, 125635]	40.2	38.7	41.4
## 2462 (54545.6, 61494.5]	32.4	32.0	32.9
## 2463 (42724.4, 45201]	42.5	41.5	43.7
## 2464 (48021.6, 51046.4]	51.3	50.5	52.6
## 2465 (48021.6, 51046.4]	42.8	42.0	44.2
## 2466 (48021.6, 51046.4]	43.5	42.7	44.5
## 2467 (61494.5, 125635]	37.5	36.5	38.6
## 2468 (61494.5, 125635]	37.9	36.9	39.0
## 2469 (51046.4, 54545.6]	44.1	42.4	45.9
## 2470 (48021.6, 51046.4]	46.7	44.1	48.3
## 2471 (51046.4, 54545.6]	43.0	41.7	44.4
## 2472 (45201, 48021.6]	44.6	43.1	46.4
## 2473 (54545.6, 61494.5]	43.0	41.3	44.4
## 2474 (48021.6, 51046.4]	46.3	44.5	48.2
## 2475 (61494.5, 125635]	36.1	35.3	37.2
## 2476 (54545.6, 61494.5]	45.0	43.6	46.4
## 2477 (45201, 48021.6]	48.0	47.7	48.4
## 2478 (61494.5, 125635]	39.5	37.8	40.4
## 2479 (48021.6, 51046.4]	45.9	44.6	47.1
## 2480 (51046.4, 54545.6]	43.4	42.7	44.3
## 2481 (48021.6, 51046.4]	43.9	42.9	44.5
## 2482 (54545.6, 61494.5]	39.9	37.7	42.2

## 2483 (48021.6, 51046.4]	48.1	46.7	49.6
## 2484 (42724.4, 45201]	47.4	46.9	47.9
## 2485 (48021.6, 51046.4]	49.8	47.8	51.6
## 2486 (51046.4, 54545.6]	50.2	50.0	50.5
## 2487 (42724.4, 45201]	50.3	51.3	48.7
## 2488 (61494.5, 125635]	41.3	41.2	41.3
## 2489 (48021.6, 51046.4]	46.5	45.1	48.4
## 2490 (51046.4, 54545.6]	35.1	33.8	36.5
## 2491 (54545.6, 61494.5]	39.9	38.7	42.0
## 2492 (37413.8, 40362.7]	36.3	35.6	37.3
## 2493 (51046.4, 54545.6]	43.8	43.2	44.3
## 2494 (51046.4, 54545.6]	45.4	44.4	47.6
## 2495 (51046.4, 54545.6]	42.0	41.8	42.2
## 2496 (48021.6, 51046.4]	40.3	38.9	41.7
## 2497 (48021.6, 51046.4]	41.2	40.0	42.5
## 2498 (45201, 48021.6]	39.0	37.7	40.8
## 2499 (51046.4, 54545.6]	47.6	46.7	48.5
## 2500 (54545.6, 61494.5]	35.7	35.6	35.7
## 2501 (48021.6, 51046.4]	36.6	35.6	37.6
## 2502 (51046.4, 54545.6]	45.4	44.4	47.1
## 2503 (61494.5, 125635]	36.7	35.8	37.7
## 2504 (51046.4, 54545.6]	46.7	45.4	48.0
## 2505 (48021.6, 51046.4]	39.1	36.6	40.3
## 2506 (40362.7, 42724.4]	43.7	42.3	45.3
## 2507 (45201, 48021.6]	42.3	39.9	44.5
## 2508 (48021.6, 51046.4]	38.9	36.9	41.5
## 2509 (51046.4, 54545.6]	46.5	44.4	48.6
## 2510 (54545.6, 61494.5]	34.6	33.5	35.7
## 2511 (48021.6, 51046.4]	42.3	41.5	42.9
## 2512 (48021.6, 51046.4]	43.2	41.7	45.0
## 2513 (51046.4, 54545.6]	44.5	43.5	45.5
## 2514 (54545.6, 61494.5]	36.4	35.0	37.4
## 2515 (54545.6, 61494.5]	41.1	40.3	42.4
## 2516 (51046.4, 54545.6]	41.2	41.1	41.3
## 2517 (48021.6, 51046.4]	40.9	39.4	42.7
## 2518 (61494.5, 125635]	35.5	35.2	35.9
## 2519 (61494.5, 125635]	35.3	34.9	35.7
## 2520 (54545.6, 61494.5]	41.4	40.1	42.7
## 2521 (54545.6, 61494.5]	34.0	33.0	34.8
## 2522 (54545.6, 61494.5]	39.8	39.1	40.2
## 2523 (54545.6, 61494.5]	32.7	31.4	33.9
## 2524 (48021.6, 51046.4]	44.5	43.5	45.8
## 2525 (42724.4, 45201]	43.0	42.1	44.0
## 2526 (48021.6, 51046.4]	48.4	47.1	49.5
## 2527 (54545.6, 61494.5]	43.8	43.1	44.7
## 2528 (40362.7, 42724.4]	43.3	41.6	44.7
## 2529 (51046.4, 54545.6]	39.0	38.2	39.6
## 2530 (61494.5, 125635]	38.9	37.8	40.0
## 2531 (48021.6, 51046.4]	39.6	38.1	41.2
## 2532 (51046.4, 54545.6]	43.7	41.9	45.7
## 2533 (48021.6, 51046.4]	33.9	33.8	33.9
## 2534 (61494.5, 125635]	35.6	35.3	36.0
## 2535 (48021.6, 51046.4]	43.0	40.7	45.1
## 3019 (37413.8, 40362.7]	41.2	39.9	43.4

## 3020	(45201, 48021.6]	41.7	40.4	42.7
## 3021	(42724.4, 45201]	35.9	34.1	37.4
## 3022	(51046.4, 54545.6]	44.9	45.3	44.7
## 3023	(45201, 48021.6]	39.3	38.5	40.5
## 3024	(37413.8, 40362.7]	37.3	34.0	40.7
## 3025	(45201, 48021.6]	42.5	41.4	43.5
## 3026	(54545.6, 61494.5]	37.8	36.5	38.7
## 3027	(45201, 48021.6]	46.2	45.0	47.6
## 3028	(37413.8, 40362.7]	49.1	46.8	51.0
## 3029	(40362.7, 42724.4]	49.6	46.4	52.5
## 3030	(45201, 48021.6]	44.4	42.2	44.8
## 3031	(51046.4, 54545.6]	42.2	41.1	44.1
## 3032	(40362.7, 42724.4]	42.3	40.1	45.5
## 3033	(54545.6, 61494.5]	44.1	42.9	44.6
## 3034	(45201, 48021.6]	45.7	43.8	49.2
## 3035	(42724.4, 45201]	38.2	36.5	40.0
## 3036	(40362.7, 42724.4]	52.3	51.9	52.8
## 3037	(45201, 48021.6]	41.7	41.2	42.3
## 3038	(45201, 48021.6]	38.8	37.3	40.5
## 3039	(48021.6, 51046.4]	28.8	28.0	29.7
## 3040	(48021.6, 51046.4]	45.2	45.2	45.3
## 3041	(34218.1, 37413.8]	50.4	49.1	52.2
## 3042	(48021.6, 51046.4]	32.2	31.0	33.8
## 3043	(45201, 48021.6]	44.2	41.1	48.8
## 3044	(48021.6, 51046.4]	30.4	29.3	31.4
## 3045	(51046.4, 54545.6]	30.9	30.5	31.2
## 3046	(48021.6, 51046.4]	39.0	36.9	40.5
## 3047	(40362.7, 42724.4]	26.2	25.5	27.0
##	Geography	AvgHouseholdSize	PercentMarried	
## 112	Churchill County, Nevada	2.5100	51.6	
## 113	Clark County, Nevada	2.7800	44.6	
## 114	Douglas County, Nevada	2.3700	58.9	
## 115	Elko County, Nevada	2.8700	56.6	
## 116	Esmeralda County, Nevada	2.3500	44.3	
## 117	Eureka County, Nevada	2.1700	64.7	
## 118	Humboldt County, Nevada	2.7400	51.6	
## 119	Lander County, Nevada	2.7900	54.3	
## 120	Lincoln County, Nevada	2.5500	56.8	
## 121	Lyon County, Nevada	2.6300	51.1	
## 122	Mineral County, Nevada	0.0225	41.8	
## 123	Nye County, Nevada	0.0242	54.4	
## 124	Pershing County, Nevada	2.2900	44.8	
## 125	Storey County, Nevada	2.2000	53.5	
## 126	Washoe County, Nevada	2.5800	47.0	
## 127	White Pine County, Nevada	2.7100	44.8	
## 128	Carson City, Nevada	2.4300	45.1	
## 149	Mercer County, New Jersey	2.6900	47.2	
## 368	Aitkin County, Minnesota	2.0500	61.2	
## 369	Anoka County, Minnesota	2.7000	54.4	
## 370	Becker County, Minnesota	2.4100	57.6	
## 371	Beltrami County, Minnesota	2.5900	44.2	
## 372	Benton County, Minnesota	2.4500	50.6	
## 568	Pulaski County, Arkansas	2.4900	43.5	
## 1014	Placer County, California	2.6700	55.6	

## 1026	Santa Barbara County, California	2.9200	44.9
## 1295	Gove County, Kansas	2.2500	69.1
## 1296	Graham County, Kansas	2.1200	60.8
## 1297	Grant County, Kansas	2.8600	62.5
## 1298	Gray County, Kansas	2.7800	64.8
## 1299	Greeley County, Kansas	2.4600	66.6
## 1300	Greenwood County, Kansas	2.2300	55.3
## 1301	Hamilton County, Kansas	2.5900	61.8
## 1302	Harper County, Kansas	2.3400	56.6
## 1303	Harvey County, Kansas	2.4700	58.4
## 1304	Haskell County, Kansas	2.9300	58.6
## 1305	Hodgeman County, Kansas	2.3600	65.5
## 1306	Jackson County, Kansas	2.5100	61.4
## 1307	Jefferson County, Kansas	2.4700	61.7
## 1308	Jewell County, Kansas	2.1000	63.5
## 1309	Johnson County, Kansas	2.5600	56.8
## 1310	Kearny County, Kansas	2.8100	63.0
## 1311	Kingman County, Kansas	2.5400	56.6
## 1312	Kiowa County, Kansas	2.2100	60.0
## 1313	Lane County, Kansas	2.0700	55.9
## 1314	Leavenworth County, Kansas	2.7000	53.1
## 1315	Lincoln County, Kansas	2.3100	56.3
## 1316	Linn County, Kansas	2.2600	61.7
## 1317	Logan County, Kansas	2.2100	59.7
## 1318	Lyon County, Kansas	2.4300	46.0
## 1319	McPherson County, Kansas	2.3700	58.9
## 1320	Marion County, Kansas	2.4000	57.9
## 1321	Marshall County, Kansas	2.2700	61.4
## 1322	Meade County, Kansas	2.5100	61.7
## 1323	Miami County, Kansas	2.5500	59.3
## 1324	Mitchell County, Kansas	2.2500	59.8
## 1325	Montgomery County, Kansas	2.4700	50.3
## 1326	Morris County, Kansas	2.3600	62.0
## 1327	Morton County, Kansas	2.5300	54.2
## 1328	Nemaha County, Kansas	2.3200	61.9
## 1329	Neosho County, Kansas	2.4700	53.1
## 1330	Ness County, Kansas	2.1800	62.5
## 1331	Norton County, Kansas	2.2400	49.6
## 1332	Osage County, Kansas	2.4400	57.4
## 1333	Osborne County, Kansas	2.0800	59.6
## 1334	Ottawa County, Kansas	2.3900	63.8
## 1335	Pawnee County, Kansas	2.3200	46.2
## 1336	Phillips County, Kansas	2.2900	60.5
## 1337	Pottawatomie County, Kansas	2.7200	62.3
## 1338	Pratt County, Kansas	2.3900	55.5
## 1339	Rawlins County, Kansas	2.0700	61.7
## 1340	Reno County, Kansas	2.4400	53.3
## 1341	Republic County, Kansas	2.0700	57.4
## 1342	Rice County, Kansas	2.4200	56.4
## 1343	Riley County, Kansas	2.4900	39.1
## 1344	Rooks County, Kansas	2.2800	57.4
## 1345	Rush County, Kansas	2.1000	56.8
## 1346	Russell County, Kansas	2.1200	58.0
## 1347	Saline County, Kansas	2.4200	48.9

## 1348	Scott County, Kansas	0.0221	65.3
## 1349	Sedgwick County, Kansas	2.5900	49.8
## 1350	Seward County, Kansas	3.0500	47.7
## 1351	Shawnee County, Kansas	2.4400	49.5
## 1352	Sheridan County, Kansas	2.2100	64.1
## 1353	Sherman County, Kansas	2.1400	54.2
## 1354	Smith County, Kansas	2.2100	60.7
## 1355	Stafford County, Kansas	2.3500	63.1
## 1356	Stanton County, Kansas	2.5800	63.5
## 1357	Stevens County, Kansas	2.8400	68.0
## 1358	Sumner County, Kansas	2.5500	56.3
## 1359	Thomas County, Kansas	2.3900	55.6
## 1360	Trego County, Kansas	2.2100	66.0
## 1361	Wabaunsee County, Kansas	2.5900	61.2
## 1362	Wallace County, Kansas	2.5200	60.1
## 1363	Washington County, Kansas	2.2700	62.5
## 1364	Wichita County, Kansas	2.7900	60.6
## 1365	Wilson County, Kansas	2.3700	53.2
## 1366	Woodson County, Kansas	2.0800	53.6
## 1367	Wyandotte County, Kansas	2.7100	41.7
## 1486	Seminole County, Florida	2.8500	46.6
## 1564	Clackamas County, Oregon	2.5900	54.9
## 1744	Northampton County, Pennsylvania	2.5600	51.4
## 1840	Fort Bend County, Texas	3.1700	58.7
## 1924	Montgomery County, Texas	2.8800	56.3
## 2454	Big Stone County, Minnesota	2.1800	63.2
## 2455	Blue Earth County, Minnesota	2.4600	43.6
## 2456	Brown County, Minnesota	2.2700	55.4
## 2457	Carlton County, Minnesota	2.5100	50.6
## 2458	Carver County, Minnesota	2.7600	62.3
## 2459	Cass County, Minnesota	2.1900	61.1
## 2460	Chippewa County, Minnesota	2.3600	56.4
## 2461	Chisago County, Minnesota	2.6300	57.2
## 2462	Clay County, Minnesota	2.5100	49.6
## 2463	Clearwater County, Minnesota	2.4800	57.1
## 2464	Cook County, Minnesota	1.9300	63.3
## 2465	Cottonwood County, Minnesota	2.3800	57.7
## 2466	Crow Wing County, Minnesota	2.3600	56.3
## 2467	Dakota County, Minnesota	2.5900	55.2
## 2468	Dodge County, Minnesota	2.6700	59.6
## 2469	Douglas County, Minnesota	2.3200	59.8
## 2470	Faribault County, Minnesota	2.1800	57.4
## 2471	Fillmore County, Minnesota	2.4000	59.1
## 2472	Freeborn County, Minnesota	2.3300	52.7
## 2473	Goodhue County, Minnesota	2.4200	55.8
## 2474	Grant County, Minnesota	2.3200	60.1
## 2475	Hennepin County, Minnesota	2.3900	47.2
## 2476	Houston County, Minnesota	2.3500	58.1
## 2477	Hubbard County, Minnesota	2.3200	62.6
## 2478	Isanti County, Minnesota	2.6800	53.6
## 2479	Itasca County, Minnesota	2.3300	58.5
## 2480	Jackson County, Minnesota	2.3200	59.4
## 2481	Kanabec County, Minnesota	2.5500	54.9
## 2482	Kandiyohi County, Minnesota	2.4800	57.7

## 2483	Kittson County, Minnesota	2.2800	54.7
## 2484	Koochiching County, Minnesota	2.1800	55.0
## 2485	Lac qui Parle County, Minnesota	2.2400	61.2
## 2486	Lake County, Minnesota	2.0600	62.3
## 2487	Lake of the Woods County, Minnesota	2.4000	52.3
## 2488	Le Sueur County, Minnesota	2.5200	58.4
## 2489	Lincoln County, Minnesota	2.2500	59.1
## 2490	Lyon County, Minnesota	2.4600	51.3
## 2491	McLeod County, Minnesota	2.4000	56.0
## 2492	Mahnomen County, Minnesota	2.7100	46.3
## 2493	Marshall County, Minnesota	2.3200	59.4
## 2494	Martin County, Minnesota	2.2600	54.9
## 2495	Meeker County, Minnesota	2.4800	60.4
## 2496	Mille Lacs County, Minnesota	2.5200	51.9
## 2497	Morrison County, Minnesota	2.4300	57.5
## 2498	Mower County, Minnesota	2.5100	53.1
## 2499	Murray County, Minnesota	2.2400	61.4
## 2500	Nicollet County, Minnesota	2.3900	50.8
## 2501	Nobles County, Minnesota	2.7100	54.1
## 2502	Norman County, Minnesota	2.4000	56.2
## 2503	Olmsted County, Minnesota	2.5200	55.2
## 2504	Otter Tail County, Minnesota	2.3400	60.2
## 2505	Pennington County, Minnesota	2.3400	49.7
## 2506	Pine County, Minnesota	2.4200	51.0
## 2507	Pipestone County, Minnesota	0.0230	59.2
## 2508	Polk County, Minnesota	2.3700	52.8
## 2509	Pope County, Minnesota	2.2300	63.0
## 2510	Ramsey County, Minnesota	2.4700	44.0
## 2511	Red Lake County, Minnesota	2.4300	57.5
## 2512	Redwood County, Minnesota	2.4300	54.0
## 2513	Renville County, Minnesota	2.3400	57.1
## 2514	Rice County, Minnesota	2.5000	49.2
## 2515	Rock County, Minnesota	2.3700	61.0
## 2516	Roseau County, Minnesota	2.4400	59.7
## 2517	St. Louis County, Minnesota	2.2600	47.7
## 2518	Scott County, Minnesota	2.9000	60.3
## 2519	Sherburne County, Minnesota	2.8900	57.2
## 2520	Sibley County, Minnesota	2.4600	58.3
## 2521	Stearns County, Minnesota	2.5200	50.0
## 2522	Steele County, Minnesota	2.5200	55.0
## 2523	Stevens County, Minnesota	2.5500	51.3
## 2524	Swift County, Minnesota	2.2100	57.5
## 2525	Todd County, Minnesota	2.4500	58.7
## 2526	Traverse County, Minnesota	2.1900	62.9
## 2527	Wabasha County, Minnesota	2.3600	61.8
## 2528	Wadena County, Minnesota	2.3500	52.2
## 2529	Waseca County, Minnesota	2.4300	56.8
## 2530	Washington County, Minnesota	2.6700	57.2
## 2531	Watsonwan County, Minnesota	2.4500	55.4
## 2532	Wilkin County, Minnesota	2.2900	56.0
## 2533	Winona County, Minnesota	2.4600	45.1
## 2534	Wright County, Minnesota	2.8200	60.5
## 2535	Yellow Medicine County, Minnesota	2.3300	58.8
## 3019	Allen County, Kansas	2.3900	50.4

## 3020	Anderson County, Kansas	2.3700	59.5
## 3021	Atchison County, Kansas	2.5300	45.6
## 3022	Barber County, Kansas	2.3800	61.0
## 3023	Barton County, Kansas	2.3800	54.3
## 3024	Bourbon County, Kansas	2.5400	53.2
## 3025	Brown County, Kansas	2.3800	56.5
## 3026	Butler County, Kansas	2.6300	56.2
## 3027	Chase County, Kansas	2.3400	54.4
## 3028	Chautauqua County, Kansas	2.2100	59.8
## 3029	Cheyenne County, Kansas	2.0900	61.8
## 3030	Clark County, Kansas	2.2500	57.7
## 3031	Clay County, Kansas	2.4500	62.3
## 3032	Cloud County, Kansas	2.2400	49.1
## 3033	Coffey County, Kansas	2.3300	58.5
## 3034	Comanche County, Kansas	2.4800	60.2
## 3035	Cowley County, Kansas	2.4800	51.9
## 3036	Decatur County, Kansas	1.9100	54.4
## 3037	Dickinson County, Kansas	2.4500	58.6
## 3038	Doniphan County, Kansas	2.3200	55.3
## 3039	Douglas County, Kansas	2.4300	39.5
## 3040	Edwards County, Kansas	2.2700	60.9
## 3041	Elk County, Kansas	2.1000	60.3
## 3042	Ellis County, Kansas	2.3500	45.1
## 3043	Ellsworth County, Kansas	2.0800	51.0
## 3044	Finney County, Kansas	2.9000	52.6
## 3045	Ford County, Kansas	3.0400	54.8
## 3046	Franklin County, Kansas	2.5600	58.8
## 3047	Geary County, Kansas	2.8300	59.5
##	PctNoHS18_24 PctHS18_24 PctSomeCol18_24 PctBachDeg18_24 PctHS25_Over		
## 112	19.8 39.3 NA 1.4 35.0		
## 113	20.2 35.5 NA 4.9 28.8		
## 114	17.3 38.6 NA 5.0 24.7		
## 115	19.0 33.7 42.3 5.1 26.8		
## 116	1.2 62.7 36.1 0.0 43.2		
## 117	48.1 19.2 32.7 0.0 35.8		
## 118	31.9 40.2 NA 4.7 37.4		
## 119	24.3 51.6 NA 3.3 37.5		
## 120	30.1 38.4 30.7 0.8 32.1		
## 121	30.2 36.7 NA 0.2 30.2		
## 122	6.1 42.3 51.6 0.0 31.2		
## 123	32.3 41.2 NA 3.3 35.6		
## 124	38.0 28.9 NA 0.0 35.4		
## 125	16.9 43.5 NA 13.7 29.5		
## 126	15.9 28.7 NA 7.8 24.0		
## 127	23.9 47.1 NA 4.3 34.5		
## 128	17.4 33.7 NA 5.1 28.6		
## 149	12.1 25.0 50.9 12.0 25.6		
## 368	19.4 42.4 NA 8.0 39.9		
## 369	15.9 32.3 NA 9.6 28.4		
## 370	18.5 38.2 36.8 6.5 33.5		
## 371	12.8 24.7 NA 6.9 26.1		
## 372	12.6 27.0 52.4 8.0 32.1		
## 568	13.0 31.0 NA 11.5 27.0		
## 1014	10.7 31.2 NA 7.4 19.4		

## 1026	11.1	19.2	NA	9.5	17.7
## 1295	8.7	29.1	NA	6.3	37.7
## 1296	19.2	21.9	NA	17.9	32.7
## 1297	37.8	33.0	NA	0.0	28.9
## 1298	33.5	25.6	NA	4.5	25.6
## 1299	4.8	0.0	NA	40.3	30.3
## 1300	8.6	39.1	45.7	6.6	40.1
## 1301	25.1	38.7	NA	8.4	34.8
## 1302	17.1	44.7	NA	6.3	36.6
## 1303	13.0	23.7	NA	11.5	29.6
## 1304	38.0	36.7	NA	4.0	20.9
## 1305	24.5	34.3	NA	2.0	25.9
## 1306	23.4	30.5	NA	7.8	40.6
## 1307	16.0	37.6	NA	4.7	40.7
## 1308	6.3	24.6	44.0	25.1	39.4
## 1309	11.5	25.0	NA	17.1	15.2
## 1310	30.0	16.6	NA	3.5	23.4
## 1311	10.3	45.8	NA	2.3	34.8
## 1312	8.5	15.0	NA	6.5	26.9
## 1313	25.2	31.1	NA	3.0	29.7
## 1314	13.3	30.5	NA	8.1	30.9
## 1315	11.2	44.1	NA	1.8	32.3
## 1316	20.6	34.7	NA	0.9	37.2
## 1317	27.5	21.3	50.2	0.9	35.5
## 1318	9.3	23.6	NA	7.7	33.4
## 1319	15.9	21.6	NA	11.3	29.1
## 1320	15.2	20.9	55.8	8.1	36.4
## 1321	16.9	28.9	47.3	6.9	44.8
## 1322	28.1	44.5	NA	1.2	31.1
## 1323	14.5	36.6	NA	5.3	32.5
## 1324	16.5	40.7	NA	4.2	34.0
## 1325	17.3	31.6	NA	3.9	32.1
## 1326	20.8	36.6	NA	6.3	35.7
## 1327	21.1	51.0	21.7	6.2	32.5
## 1328	13.6	42.0	NA	10.0	41.9
## 1329	9.4	20.7	NA	4.7	33.6
## 1330	31.1	27.4	NA	17.8	35.9
## 1331	25.1	18.9	41.3	14.7	34.6
## 1332	15.7	33.8	NA	8.8	40.5
## 1333	18.0	22.0	NA	15.6	36.3
## 1334	17.1	36.5	NA	7.3	32.9
## 1335	26.0	23.2	49.6	1.2	28.0
## 1336	33.2	13.9	45.3	7.6	35.3
## 1337	9.2	35.0	NA	6.5	31.2
## 1338	11.9	37.3	NA	3.9	26.5
## 1339	29.1	28.5	NA	2.5	30.8
## 1340	20.2	32.7	NA	2.4	28.6
## 1341	20.2	23.6	43.4	12.7	35.1
## 1342	20.5	22.2	NA	4.6	30.5
## 1343	2.5	20.3	NA	8.3	17.4
## 1344	12.6	42.1	NA	12.6	36.8
## 1345	22.0	20.6	43.0	14.5	33.0
## 1346	25.2	31.0	40.6	3.2	32.8
## 1347	14.5	29.4	NA	6.2	33.5

## 1348	21.0	18.3	48.3	12.3	33.4
## 1349	15.4	28.8	NA	8.5	26.8
## 1350	25.7	23.7	NA	2.7	24.8
## 1351	15.8	31.4	NA	7.5	31.1
## 1352	6.7	18.7	NA	8.2	33.8
## 1353	25.6	10.1	NA	10.5	33.3
## 1354	10.0	37.2	NA	4.8	35.3
## 1355	25.3	21.7	NA	10.7	28.7
## 1356	21.3	27.0	NA	3.5	27.4
## 1357	38.7	29.0	NA	14.8	35.6
## 1358	20.4	31.1	NA	5.0	35.9
## 1359	12.2	20.1	NA	3.8	28.9
## 1360	3.4	26.3	NA	14.8	33.2
## 1361	10.8	40.7	45.4	3.1	38.9
## 1362	25.6	22.2	38.9	13.3	32.9
## 1363	14.8	19.9	NA	20.4	38.0
## 1364	36.9	20.6	NA	16.3	29.6
## 1365	20.0	34.4	NA	9.0	36.8
## 1366	7.8	46.1	41.1	5.0	38.8
## 1367	24.7	33.7	NA	6.8	32.9
## 1486	11.2	24.9	53.6	10.2	24.1
## 1564	11.8	33.9	NA	9.3	23.5
## 1744	9.3	28.7	NA	9.1	35.3
## 1840	15.2	28.8	NA	8.8	17.5
## 1924	20.3	32.8	NA	7.4	24.3
## 2454	43.3	20.7	NA	6.1	42.4
## 2455	3.1	15.9	NA	7.0	28.1
## 2456	8.6	28.9	NA	5.3	37.6
## 2457	13.8	34.3	NA	3.9	36.4
## 2458	13.1	26.7	45.7	14.5	21.2
## 2459	18.0	44.2	NA	3.6	35.4
## 2460	22.7	29.7	NA	4.9	36.0
## 2461	19.1	37.8	NA	5.6	34.6
## 2462	5.2	24.4	NA	7.4	27.3
## 2463	20.8	37.4	35.6	6.1	38.1
## 2464	11.8	25.8	NA	11.5	25.1
## 2465	29.4	36.8	27.0	6.8	36.2
## 2466	12.8	29.2	NA	5.7	31.4
## 2467	13.4	30.2	43.0	13.4	21.6
## 2468	13.3	38.6	NA	7.0	33.5
## 2469	11.5	25.3	NA	9.1	29.9
## 2470	15.6	34.1	NA	6.0	39.9
## 2471	21.7	31.1	NA	5.9	35.7
## 2472	15.9	36.6	42.2	5.3	36.7
## 2473	14.8	36.9	38.9	9.4	33.0
## 2474	15.7	29.4	44.0	10.9	33.4
## 2475	12.1	22.5	NA	19.9	17.8
## 2476	11.8	30.3	NA	8.9	34.3
## 2477	19.9	42.5	NA	6.1	31.8
## 2478	13.1	42.5	NA	3.5	37.8
## 2479	14.7	27.1	52.7	5.5	31.8
## 2480	10.7	28.0	NA	7.9	34.8
## 2481	14.6	40.1	NA	4.5	39.7
## 2482	18.7	24.0	NA	8.4	27.9

## 2483	16.4	33.9	41.6	8.1	35.0
## 2484	18.8	34.7	NA	6.7	36.5
## 2485	23.3	33.7	NA	4.5	38.3
## 2486	13.3	41.9	NA	4.8	36.2
## 2487	12.9	62.5	18.8	5.8	44.8
## 2488	12.8	38.9	NA	8.7	35.5
## 2489	17.9	31.6	NA	11.3	40.5
## 2490	7.3	23.5	61.3	7.9	35.1
## 2491	15.1	27.8	NA	9.3	38.0
## 2492	27.5	41.0	26.8	4.7	37.1
## 2493	20.4	35.1	40.0	4.6	37.4
## 2494	19.6	29.7	NA	6.3	38.6
## 2495	20.2	37.2	NA	4.5	38.3
## 2496	18.4	33.4	38.6	9.6	38.3
## 2497	16.2	35.1	44.3	4.4	38.6
## 2498	27.4	23.4	NA	3.7	32.6
## 2499	18.9	32.6	NA	9.6	36.5
## 2500	3.8	14.8	NA	8.0	27.0
## 2501	41.1	29.3	24.9	4.7	34.3
## 2502	14.4	31.8	46.8	7.0	38.5
## 2503	14.0	27.3	42.1	16.6	20.9
## 2504	16.4	31.1	NA	9.5	30.5
## 2505	13.4	27.5	52.8	6.3	36.0
## 2506	14.8	40.7	NA	3.4	42.6
## 2507	22.5	31.1	NA	2.9	36.0
## 2508	14.9	26.5	NA	5.1	31.1
## 2509	20.2	32.3	NA	6.9	34.1
## 2510	12.7	25.0	NA	15.4	21.9
## 2511	12.9	34.9	43.5	8.6	40.9
## 2512	17.3	35.1	NA	10.0	37.9
## 2513	16.1	40.8	34.0	9.1	37.5
## 2514	8.8	18.6	69.0	3.7	31.8
## 2515	14.4	36.7	NA	6.7	38.5
## 2516	18.9	40.7	36.8	3.6	37.3
## 2517	9.0	21.7	59.4	9.9	29.5
## 2518	15.6	27.9	NA	14.9	23.4
## 2519	16.4	34.3	NA	7.0	26.6
## 2520	18.9	38.5	NA	4.9	40.8
## 2521	7.0	16.5	NA	7.6	30.2
## 2522	15.5	32.9	NA	9.1	35.5
## 2523	5.3	22.9	NA	5.2	27.2
## 2524	17.6	38.7	NA	6.8	38.6
## 2525	24.4	27.7	43.1	4.9	37.1
## 2526	10.3	13.3	59.1	17.2	34.8
## 2527	13.8	40.1	39.2	6.8	39.1
## 2528	22.3	27.7	46.1	3.9	39.3
## 2529	12.6	48.0	34.7	4.8	35.2
## 2530	13.6	28.9	NA	13.2	22.0
## 2531	24.0	40.1	NA	4.8	40.0
## 2532	16.1	28.6	48.2	7.0	30.5
## 2533	4.4	24.2	NA	6.9	29.6
## 2534	16.1	26.8	NA	9.2	28.9
## 2535	15.8	37.9	NA	4.9	36.8
## 3019	7.3	32.4	56.8	3.5	33.4

## 3020	12.7	30.5	50.1	6.7	39.1
## 3021	5.0	22.8	NA	8.7	41.0
## 3022	26.2	24.4	NA	4.3	36.6
## 3023	13.8	37.4	NA	4.4	33.9
## 3024	12.6	28.0	NA	3.9	31.2
## 3025	21.4	28.3	NA	13.5	39.8
## 3026	15.8	29.0	NA	7.3	26.8
## 3027	15.9	47.7	NA	6.8	33.5
## 3028	19.1	42.7	36.5	1.7	36.1
## 3029	21.4	18.5	NA	12.7	31.5
## 3030	19.6	33.1	NA	12.8	26.1
## 3031	7.7	26.5	50.5	15.2	33.7
## 3032	14.8	25.3	NA	2.7	34.2
## 3033	16.1	31.8	NA	5.9	36.8
## 3034	32.2	36.5	26.1	5.2	28.4
## 3035	12.8	32.3	46.0	8.9	30.5
## 3036	11.3	40.6	NA	18.1	37.6
## 3037	12.4	44.1	NA	4.7	36.0
## 3038	7.3	31.7	NA	4.1	38.7
## 3039	2.8	13.1	NA	13.9	20.0
## 3040	16.3	29.2	NA	14.4	30.7
## 3041	6.5	59.7	NA	4.8	35.3
## 3042	2.9	17.7	NA	9.8	27.7
## 3043	20.9	27.9	NA	8.4	32.2
## 3044	26.7	33.9	35.6	3.8	23.1
## 3045	19.7	44.5	33.3	2.5	23.0
## 3046	10.9	37.2	NA	4.1	36.1
## 3047	8.0	43.3	43.9	4.8	29.8
##	PctBachDeg25_Over	PctEmployed16_Over	PctUnemployed16_Over		
## 112	10.9	48.4	11.9		
## 113	15.1	57.7	10.8		
## 114	16.4	51.5	8.5		
## 115	12.3	67.0	5.5		
## 116	12.6	46.0	11.9		
## 117	14.2	64.0	4.7		
## 118	8.8	64.5	7.8		
## 119	4.7	59.8	12.0		
## 120	15.9	41.0	11.8		
## 121	11.1	48.6	13.6		
## 122	6.8	48.1	15.4		
## 123	9.2	39.7	13.4		
## 124	7.5	36.9	9.5		
## 125	14.1	50.4	10.3		
## 126	18.2	59.6	9.1		
## 127	8.8	43.8	10.6		
## 128	12.8	52.9	11.3		
## 149	20.7	59.5	9.4		
## 368	10.8	47.8	7.3		
## 369	20.0	68.4	6.2		
## 370	15.6	61.5	3.5		
## 371	17.4	58.5	10.6		
## 372	15.2	66.0	6.7		
## 568	20.7	NA	7.5		
## 1014	24.5	55.8	8.3		

## 1026	19.2	58.4	8.0
## 1295	17.4	62.6	1.9
## 1296	20.0	59.6	2.4
## 1297	14.0	64.7	7.3
## 1298	14.3	70.8	2.9
## 1299	20.4	60.5	2.1
## 1300	11.3	56.2	3.5
## 1301	11.1	67.4	2.3
## 1302	13.0	59.7	3.7
## 1303	17.3	60.8	4.4
## 1304	11.2	68.2	3.9
## 1305	18.4	65.0	2.5
## 1306	12.5	62.8	4.4
## 1307	12.7	62.0	6.9
## 1308	11.3	58.3	2.5
## 1309	33.6	69.2	4.5
## 1310	10.5	66.3	1.8
## 1311	14.8	56.5	5.7
## 1312	17.9	NA	4.0
## 1313	15.5	67.9	2.4
## 1314	16.4	54.7	5.9
## 1315	15.6	59.5	6.5
## 1316	10.5	52.1	8.3
## 1317	11.7	63.0	3.9
## 1318	16.6	62.7	7.2
## 1319	19.3	66.0	3.1
## 1320	14.7	58.4	5.0
## 1321	11.8	62.1	3.0
## 1322	16.4	NA	2.8
## 1323	16.5	63.4	5.6
## 1324	18.0	60.5	2.6
## 1325	11.2	53.8	8.7
## 1326	10.0	59.9	3.2
## 1327	10.6	57.1	5.8
## 1328	16.8	67.1	1.7
## 1329	12.1	60.9	4.7
## 1330	15.5	60.9	2.7
## 1331	15.2	57.0	1.7
## 1332	12.8	57.3	8.2
## 1333	15.1	61.1	4.0
## 1334	15.2	63.2	5.3
## 1335	12.7	51.6	1.9
## 1336	15.5	63.1	4.3
## 1337	21.2	64.8	3.7
## 1338	18.0	59.9	5.6
## 1339	18.6	58.2	2.3
## 1340	13.8	58.2	5.7
## 1341	13.4	61.8	3.8
## 1342	14.0	61.9	5.5
## 1343	24.1	54.5	5.8
## 1344	14.1	63.2	4.1
## 1345	12.8	59.7	5.6
## 1346	13.1	60.7	6.9
## 1347	15.2	65.0	5.0

## 1348	15.7	67.5	5.8
## 1349	19.7	61.9	6.9
## 1350	8.5	66.2	9.0
## 1351	18.7	60.2	6.8
## 1352	15.9	65.1	1.2
## 1353	9.8	64.3	4.0
## 1354	13.0	59.4	3.2
## 1355	17.0	62.1	2.6
## 1356	16.4	71.0	2.4
## 1357	12.7	64.4	3.0
## 1358	14.1	58.2	7.0
## 1359	18.6	68.9	2.7
## 1360	18.5	69.6	1.0
## 1361	16.1	60.9	6.2
## 1362	19.6	NA	5.4
## 1363	13.2	64.5	3.1
## 1364	12.5	67.4	3.1
## 1365	9.9	56.2	6.5
## 1366	12.7	59.7	4.6
## 1367	10.7	58.7	11.2
## 1486	23.6	59.4	9.4
## 1564	21.5	59.2	8.3
## 1744	17.2	59.2	7.3
## 1840	28.0	63.7	5.1
## 1924	21.7	NA	6.1
## 2454	13.4	57.6	3.3
## 2455	21.6	68.0	4.9
## 2456	15.3	66.2	3.2
## 2457	16.8	58.7	5.4
## 2458	32.0	73.2	3.8
## 2459	14.2	52.6	7.4
## 2460	11.8	65.6	2.6
## 2461	15.6	64.7	5.5
## 2462	22.8	68.8	4.3
## 2463	11.1	53.2	7.4
## 2464	22.3	62.0	4.0
## 2465	13.9	59.2	4.9
## 2466	15.6	57.9	5.7
## 2467	27.8	NA	5.0
## 2468	18.1	71.4	3.5
## 2469	17.8	63.8	4.5
## 2470	12.8	61.5	4.2
## 2471	13.8	64.5	4.3
## 2472	11.5	61.5	4.9
## 2473	16.8	63.9	5.3
## 2474	14.6	61.1	3.6
## 2475	30.3	67.9	6.0
## 2476	14.9	64.4	4.9
## 2477	16.7	56.6	5.0
## 2478	12.8	NA	7.3
## 2479	15.1	54.2	7.1
## 2480	14.2	66.0	3.2
## 2481	8.0	56.3	9.8
## 2482	15.1	64.9	4.8

## 2483	16.8	60.3	2.8
## 2484	13.5	54.4	7.8
## 2485	13.6	60.8	3.7
## 2486	15.5	54.0	4.8
## 2487	12.4	59.9	4.8
## 2488	15.4	66.9	5.3
## 2489	15.8	63.9	2.8
## 2490	18.6	69.0	5.3
## 2491	12.9	66.7	4.5
## 2492	9.8	56.2	9.2
## 2493	14.5	62.8	4.7
## 2494	15.1	62.5	3.5
## 2495	13.6	63.6	4.7
## 2496	10.0	58.4	7.7
## 2497	11.6	63.6	5.8
## 2498	13.8	62.5	6.5
## 2499	13.0	61.1	4.5
## 2500	22.3	70.4	4.0
## 2501	10.1	63.0	6.9
## 2502	12.4	57.9	5.1
## 2503	23.8	69.3	4.1
## 2504	17.2	59.3	4.4
## 2505	11.8	68.5	3.0
## 2506	9.1	53.0	7.4
## 2507	14.0	62.8	4.2
## 2508	16.3	63.9	4.6
## 2509	16.1	60.7	4.1
## 2510	24.4	63.9	7.1
## 2511	11.8	65.8	3.5
## 2512	13.0	61.8	4.0
## 2513	12.2	62.1	5.0
## 2514	17.6	NA	5.6
## 2515	15.0	63.5	3.7
## 2516	14.2	69.1	2.7
## 2517	18.2	NA	6.9
## 2518	28.1	73.2	4.6
## 2519	19.2	69.6	5.2
## 2520	12.8	66.8	5.5
## 2521	17.7	68.8	5.0
## 2522	18.7	65.2	4.9
## 2523	20.1	63.6	3.1
## 2524	12.6	63.2	3.9
## 2525	11.0	59.6	5.1
## 2526	13.3	57.4	2.9
## 2527	14.6	NA	3.6
## 2528	8.8	54.2	5.2
## 2529	15.4	63.4	4.6
## 2530	27.3	NA	4.8
## 2531	12.5	66.1	2.9
## 2532	13.5	64.6	2.0
## 2533	18.5	66.0	6.2
## 2534	20.4	71.2	5.2
## 2535	12.1	63.6	3.4
## 3019	12.2	56.7	7.0

## 3020	10.5	59.0	5.9
## 3021	12.7	57.3	7.3
## 3022	11.7	62.1	3.1
## 3023	12.4	61.4	6.8
## 3024	13.9	57.0	3.3
## 3025	12.1	60.3	4.5
## 3026	18.6	61.6	5.6
## 3027	16.5	54.1	2.9
## 3028	13.2	52.7	5.1
## 3029	16.9	59.7	1.9
## 3030	23.5	61.8	1.4
## 3031	16.0	63.2	1.9
## 3032	10.6	58.9	6.8
## 3033	15.6	62.1	5.1
## 3034	17.2	63.2	5.4
## 3035	12.7	55.4	7.0
## 3036	18.5	57.1	3.8
## 3037	13.6	58.1	6.7
## 3038	11.3	59.5	5.8
## 3039	26.3	NA	6.0
## 3040	15.0	61.3	2.0
## 3041	11.8	50.7	5.3
## 3042	21.9	68.6	4.5
## 3043	15.2	51.7	4.3
## 3044	12.4	70.1	4.6
## 3045	12.8	64.8	6.4
## 3046	14.4	NA	6.9
## 3047	13.7	48.8	9.2
##	PctPrivateCoverage	PctEmpPrivCoverage	PctPublicCoverage PctWhite
## 112	63.0	34.9	37.0 81.84067
## 113	62.9	45.4	27.3 63.75048
## 114	70.7	39.6	35.8 89.15127
## 115	73.0	60.9	18.5 88.13661
## 116	56.5	34.9	37.1 87.37949
## 117	76.4	52.6	18.6 95.80587
## 118	69.4	53.9	20.6 87.06275
## 119	69.1	53.7	24.4 88.44601
## 120	64.1	48.5	31.5 90.64305
## 121	61.1	38.6	37.5 85.73669
## 122	58.1	36.4	43.2 66.68857
## 123	53.1	30.2	48.2 85.66334
## 124	59.9	38.1	34.7 86.35823
## 125	68.5	40.6	34.1 94.24790
## 126	65.4	45.7	28.1 80.45764
## 127	67.6	48.4	31.9 86.56507
## 128	61.3	41.5	35.6 82.83653
## 149	73.1	55.9	27.4 62.92611
## 368	66.3	32.5	51.5 95.07545
## 369	79.3	60.6	25.7 85.66465
## 370	66.7	41.0	38.4 87.75726
## 371	58.6	39.6	40.5 74.08989
## 372	71.5	51.3	32.2 94.31427
## 568	64.2	43.6	33.7 58.62655
## 1014	78.2	53.6	27.6 83.64393

## 1026	63.2	39.7	31.9 74.28381
## 1295	74.7	35.1	32.3 96.89214
## 1296	77.0	37.2	38.8 91.90127
## 1297	62.3	40.6	25.6 75.42221
## 1298	76.1	48.4	22.1 98.35120
## 1299	81.9	42.7	28.8 87.17320
## 1300	69.4	40.5	38.0 95.36994
## 1301	58.9	35.7	31.1 87.97370
## 1302	71.7	44.5	32.8 96.21225
## 1303	75.1	51.2	29.9 92.91517
## 1304	62.4	41.4	27.5 85.76271
## 1305	79.2	47.0	28.2 96.02888
## 1306	77.7	52.6	31.0 87.61940
## 1307	77.6	54.2	29.1 95.68737
## 1308	74.8	31.9	41.5 97.13627
## 1309	84.0	63.0	18.9 86.91211
## 1310	77.5	54.8	21.4 83.65385
## 1311	71.5	43.3	32.0 97.18870
## 1312	73.1	43.7	34.5 95.46530
## 1313	78.2	48.6	29.2 98.32736
## 1314	80.2	42.7	26.7 83.88536
## 1315	70.8	40.8	33.0 97.52695
## 1316	75.0	44.8	32.3 96.30563
## 1317	79.1	44.8	33.5 95.41711
## 1318	67.3	46.4	29.5 88.18660
## 1319	80.7	56.0	27.8 94.78326
## 1320	74.9	49.1	33.4 95.76892
## 1321	77.7	48.1	32.6 97.08146
## 1322	76.1	49.8	27.2 88.31850
## 1323	81.1	56.0	24.5 95.58248
## 1324	79.5	47.3	27.6 97.87166
## 1325	64.2	42.8	36.8 85.46396
## 1326	75.9	42.4	34.6 96.29565
## 1327	70.9	49.6	30.1 85.83387
## 1328	80.4	50.2	28.0 97.19461
## 1329	67.8	46.2	35.0 94.60513
## 1330	77.2	42.6	35.5 96.88008
## 1331	72.8	45.9	32.9 92.20036
## 1332	71.9	44.9	35.2 96.31841
## 1333	74.6	38.9	34.8 97.67319
## 1334	77.8	50.8	27.9 96.80886
## 1335	72.8	42.1	32.8 90.65258
## 1336	75.8	44.5	33.4 96.85512
## 1337	80.4	52.1	23.8 94.08619
## 1338	71.8	47.0	32.1 95.03029
## 1339	77.6	40.8	35.1 97.40668
## 1340	69.7	46.3	32.5 91.39686
## 1341	74.7	42.0	36.9 97.66958
## 1342	70.1	47.4	30.5 90.37348
## 1343	85.4	50.6	16.4 83.26091
## 1344	71.5	40.8	35.5 96.81836
## 1345	75.2	45.4	33.5 98.14698
## 1346	70.5	34.1	39.3 95.43503
## 1347	70.9	47.3	29.1 87.64152

## 1348	84.1	50.8	24.7 95.23133
## 1349	69.2	50.3	28.0 79.56247
## 1350	61.5	49.0	26.0 59.37957
## 1351	71.7	48.5	31.9 82.79565
## 1352	73.4	31.6	37.6 97.90597
## 1353	65.3	39.5	36.7 92.76511
## 1354	76.0	31.2	39.1 97.48663
## 1355	64.7	38.9	36.8 95.57870
## 1356	76.6	52.0	23.8 75.10470
## 1357	69.7	45.1	25.2 84.85793
## 1358	72.7	50.5	30.8 94.37770
## 1359	81.0	48.3	24.7 94.47319
## 1360	83.3	47.2	26.8 95.62860
## 1361	80.9	55.2	25.7 96.39846
## 1362	75.3	40.2	26.8 99.49843
## 1363	74.4	44.7	33.9 97.22125
## 1364	71.7	32.8	33.2 94.32849
## 1365	60.1	37.3	40.0 94.62532
## 1366	63.1	35.8	38.5 94.67456
## 1367	51.4	36.8	35.3 61.71101
## 1486	67.8	47.7	26.3 79.60951
## 1564	75.3	51.8	28.3 89.27146
## 1744	77.1	54.2	30.1 86.47569
## 1840	74.0	60.0	17.3 52.50672
## 1924	68.4	51.8	23.3 86.40591
## 2454	76.1	35.3	38.7 98.03272
## 2455	79.5	56.3	26.5 92.06910
## 2456	81.5	52.1	30.6 97.36915
## 2457	73.1	53.1	34.1 89.65381
## 2458	87.5	68.6	17.0 93.04707
## 2459	61.7	31.9	49.5 85.19233
## 2460	68.5	40.2	39.5 93.87856
## 2461	80.1	58.9	27.0 95.46012
## 2462	78.8	57.8	26.5 92.60500
## 2463	57.6	34.5	45.2 86.76874
## 2464	73.9	42.9	38.6 86.98498
## 2465	70.6	36.5	43.5 90.85282
## 2466	70.1	42.4	41.1 96.52963
## 2467	82.2	63.7	22.7 84.13195
## 2468	81.8	60.1	24.7 97.25481
## 2469	77.9	46.3	35.2 97.72256
## 2470	72.1	42.1	39.6 97.65987
## 2471	74.8	46.5	33.2 98.26321
## 2472	70.3	43.6	39.0 95.12574
## 2473	77.2	51.6	31.3 95.17002
## 2474	73.5	38.6	41.2 97.53397
## 2475	75.4	56.1	27.3 74.27382
## 2476	76.7	50.6	34.2 97.24644
## 2477	67.9	36.7	44.9 94.52707
## 2478	75.0	54.7	31.1 95.76457
## 2479	69.6	43.2	40.6 93.05905
## 2480	80.4	47.7	32.2 95.34815
## 2481	64.6	39.9	44.5 96.78810
## 2482	70.7	44.6	36.1 90.42739

## 2483	73.8	41.5	36.5 98.10268
## 2484	62.3	37.3	47.5 93.71840
## 2485	78.6	38.5	38.4 96.95287
## 2486	71.9	44.3	42.5 97.03256
## 2487	69.4	41.1	40.1 94.27703
## 2488	78.7	56.1	29.6 95.36579
## 2489	78.6	45.0	33.7 97.21074
## 2490	76.7	51.4	29.5 89.67664
## 2491	77.3	52.5	32.1 95.90246
## 2492	47.9	27.7	48.3 48.76274
## 2493	76.8	46.4	32.1 97.56691
## 2494	72.5	42.8	38.8 97.57740
## 2495	75.7	49.8	32.6 97.32803
## 2496	69.7	44.7	38.4 91.26661
## 2497	71.3	44.5	37.6 97.19981
## 2498	73.4	49.0	35.0 92.21710
## 2499	76.2	40.6	36.3 96.72881
## 2500	80.7	57.5	28.0 93.60757
## 2501	67.3	42.5	33.3 80.98400
## 2502	70.3	39.2	40.5 93.21578
## 2503	80.4	60.9	26.4 85.58856
## 2504	74.7	44.1	38.3 95.89644
## 2505	75.5	52.1	33.9 93.75620
## 2506	68.4	42.1	40.4 91.61134
## 2507	69.1	37.0	37.8 93.60701
## 2508	72.5	46.4	34.5 92.85194
## 2509	76.8	43.2	37.6 97.52466
## 2510	68.8	50.6	33.6 69.12484
## 2511	72.9	49.0	36.0 94.97661
## 2512	71.6	39.4	36.5 89.44222
## 2513	76.2	43.1	35.0 94.04785
## 2514	78.4	54.9	25.9 89.99322
## 2515	73.8	44.4	34.2 96.60148
## 2516	80.4	55.1	29.6 93.61511
## 2517	74.0	49.1	35.3 92.66955
## 2518	85.5	68.0	16.8 85.67236
## 2519	82.3	64.5	22.3 93.96467
## 2520	76.6	50.3	32.2 96.77784
## 2521	77.9	54.7	28.6 91.04165
## 2522	76.1	54.6	32.1 91.33149
## 2523	84.2	52.3	27.8 93.11227
## 2524	72.0	42.4	38.6 95.91601
## 2525	66.1	41.1	37.3 95.40996
## 2526	75.8	38.4	41.7 92.44678
## 2527	77.8	49.0	34.1 97.56793
## 2528	62.5	34.6	47.7 96.39509
## 2529	77.0	52.0	31.6 93.69889
## 2530	85.7	65.1	21.2 87.00855
## 2531	69.4	41.7	34.5 94.13787
## 2532	79.1	49.2	30.4 96.70549
## 2533	78.2	51.5	26.7 94.12063
## 2534	81.3	62.1	23.3 94.56372
## 2535	74.5	42.0	35.0 92.58819
## 3019	66.0	42.7	37.7 93.46380

## 3020	67.7	33.9	37.6	95.14944
## 3021	69.3	46.0	32.0	90.57296
## 3022	73.8	40.3	34.3	95.95388
## 3023	66.2	42.4	34.0	91.89021
## 3024	65.8	43.8	37.0	93.07994
## 3025	66.7	41.0	37.9	86.00202
## 3026	76.6	55.1	26.5	93.56654
## 3027	72.9	44.3	35.3	95.77517
## 3028	62.5	34.2	43.2	88.38068
## 3029	79.7	39.3	33.5	94.70544
## 3030	75.3	52.4	29.5	91.65501
## 3031	80.2	43.3	33.8	96.71060
## 3032	71.6	41.7	34.4	94.96734
## 3033	73.0	50.2	32.1	96.16931
## 3034	74.6	43.9	32.2	97.96020
## 3035	69.0	46.2	34.4	87.71862
## 3036	67.8	35.6	40.0	96.52324
## 3037	74.0	42.3	34.5	94.50195
## 3038	69.8	45.7	29.5	91.72923
## 3039	79.3	55.3	19.6	84.29288
## 3040	68.1	39.4	30.6	93.78134
## 3041	66.0	36.2	45.5	94.37197
## 3042	80.0	52.9	24.3	94.29173
## 3043	78.3	44.6	31.7	90.28081
## 3044	64.5	48.6	28.8	75.70625
## 3045	62.0	47.8	26.6	87.96163
## 3046	75.9	49.6	29.5	92.90568
## 3047	76.1	22.9	25.1	70.09813
##	PctBlack	PctAsian	PctOtherRace	PctMarriedHouseholds BirthRate
## 112	2.37093848	2.66782121	3.64506020	49.69921 7.724034
## 113	10.83685568	9.16975671	10.14589511	43.81307 5.125073
## 114	0.55439176	1.30557143	3.49986246	56.13530 5.193754
## 115	1.28970948	0.92703929	2.56196424	57.82098 6.700734
## 116	0.00000000	0.00000000	7.97546012	40.57971 4.326923
## 117	0.11983223	0.00000000	1.43798682	56.06258 2.412869
## 118	0.45116306	0.31640007	4.37100838	51.35794 5.517423
## 119	0.13454423	0.80726539	2.37134208	56.90634 11.631799
## 120	2.92645360	0.07701194	0.90489026	59.21402 4.519774
## 121	0.93694949	1.30669609	5.42617651	51.96681 4.629630
## 122	1.31406045	4.09548839	4.35830048	34.42211 9.518477
## 123	2.56422287	1.43108504	5.15894428	52.18339 6.302577
## 124	4.81999405	0.16364177	2.91579887	51.41623 14.621131
## 125	0.43268007	1.45075083	0.20361415	52.23543 1.178203
## 126	2.40633168	5.30988302	5.80572343	45.20184 5.642579
## 127	4.17084419	0.74192901	0.92239823	53.24757 5.470460
## 128	0.96545648	2.98814287	7.25010095	42.93785 4.262413
## 149	20.14710490	10.07800936	4.85532614	49.81309 4.725836
## 368	0.36618473	0.30936296	0.08207589	53.34472 7.867412
## 369	4.90134725	4.08927749	1.69409973	56.43372 6.191393
## 370	0.51602390	0.46472328	0.37419277	54.03333 5.235358
## 371	0.81877008	0.87159396	0.26191839	44.60243 6.561705
## 372	2.32528492	1.13204661	0.11728411	48.05527 7.207015
## 568	35.70146211	2.08854616	1.11278149	40.67557 5.116648
## 1014	1.41503768	6.68013542	2.91853227	56.72986 4.433235

## 1026	2.00229437	5.16301480	13.26786739	48.71595	4.603274
## 1295	0.98720292	0.54844607	0.14625229	62.43697	8.951965
## 1296	3.08522946	1.19552642	0.00000000	53.28893	10.810811
## 1297	0.01279427	0.00000000	22.14687820	64.27780	5.375140
## 1298	0.00000000	0.13190437	1.07172300	64.88372	6.199262
## 1299	0.89869281	0.32679739	8.90522876	64.68172	5.687204
## 1300	0.15642109	0.04692633	0.86031597	49.75212	5.457746
## 1301	2.08816705	0.00000000	7.73395205	57.01403	6.605505
## 1302	0.15355741	0.05118580	0.58010578	51.95232	7.601185
## 1303	1.63628534	0.68035022	1.80565523	57.58044	5.858531
## 1304	0.00000000	0.94430993	10.12106538	63.94265	5.965293
## 1305	0.00000000	2.37235688	0.00000000	61.28237	9.393939
## 1306	0.67164179	0.48507463	0.04477612	58.75781	5.896980
## 1307	0.47624087	0.02116626	0.13228913	60.96526	5.155718
## 1308	0.06583279	0.59249506	0.49374588	54.78018	6.681034
## 1309	4.48877410	4.46019329	0.91829066	55.46135	5.529393
## 1310	0.86032389	0.27834008	10.95647773	64.97462	5.231144
## 1311	0.14120668	0.00000000	0.07702182	55.25615	7.676768
## 1312	0.86750789	0.67034700	0.51261830	56.00375	3.783784
## 1313	0.35842294	0.00000000	0.00000000	47.82609	10.029499
## 1314	8.46255129	1.57490381	1.12748795	56.92975	5.174511
## 1315	0.82435003	0.34876347	0.00000000	49.88764	7.180385
## 1316	0.69269521	0.34634761	0.15743073	54.47154	7.603802
## 1317	0.00000000	0.17901898	0.46544934	51.64924	7.558140
## 1318	2.53720638	2.04112127	3.05719921	44.49216	7.999081
## 1319	1.04608232	0.68029536	0.99480377	56.52616	6.078626
## 1320	0.81366965	0.18714402	0.17087063	57.88270	6.021595
## 1321	0.28985507	0.28985507	0.38980510	55.71495	6.352683
## 1322	0.84417066	1.04950947	8.16792151	61.09799	12.219731
## 1323	1.33688204	0.37934410	0.21720509	56.81529	6.683134
## 1324	0.38119441	0.69885642	0.14294790	55.30331	2.550091
## 1325	5.11642874	0.91855839	0.43587643	47.39859	6.411140
## 1326	0.15652174	0.40000000	0.71304348	59.43475	5.611776
## 1327	0.93367676	0.16097875	7.24404379	52.56518	3.291536
## 1328	0.88591397	0.24608721	0.14765233	56.80585	5.558363
## 1329	1.16909213	0.19484869	0.44449857	52.01919	8.276060
## 1330	0.51998700	0.00000000	1.52746181	57.19450	9.725159
## 1331	2.45080501	0.35778175	1.21645796	53.79896	9.776833
## 1332	0.14925373	0.16791045	0.16791045	55.15757	4.447806
## 1333	0.39661555	1.05764146	0.10576415	50.90703	4.141104
## 1334	0.29761905	0.00000000	0.41335979	58.52387	1.470588
## 1335	4.84734481	0.00000000	1.62060483	45.76803	1.121657
## 1336	0.38174877	0.18178513	0.16360662	55.04470	9.342231
## 1337	1.29060774	0.90165746	0.68508287	62.60986	8.095893
## 1338	0.71875963	0.01026800	0.60581169	52.74218	5.278450
## 1339	0.35363458	0.11787819	0.47151277	53.38284	10.287081
## 1340	2.98011177	0.56511287	1.67660558	52.29310	5.263956
## 1341	0.14565127	0.58260508	0.00000000	48.87912	7.552083
## 1342	1.80746954	0.48931496	4.05432395	55.39809	5.697731
## 1343	6.55274453	4.50934393	1.11167391	43.09140	4.808995
## 1344	0.23139221	0.57848052	0.55919784	51.17860	5.273834
## 1345	0.00000000	0.00000000	0.06281407	49.39435	10.758377
## 1346	1.24499141	0.00000000	0.01431025	52.20747	10.703125
## 1347	3.20983224	2.48138513	3.19368440	46.41804	6.843610

## 1348	0.00000000	1.05519480	2.82061688	56.83060	6.629834
## 1349	9.22197939	4.26273718	2.15051853	47.48265	6.454934
## 1350	3.74667010	2.94749506	29.93039443	46.64439	6.872294
## 1351	7.87395409	1.13651617	1.92402345	46.64696	5.852417
## 1352	0.07902015	0.00000000	0.15804030	56.66372	7.382550
## 1353	1.43706640	0.00000000	3.76610506	43.57502	7.904412
## 1354	0.42780749	0.16042781	0.05347594	56.26875	8.463950
## 1355	0.97222222	0.43981482	1.57407407	59.25307	6.165919
## 1356	1.02373197	0.60493253	17.21731038	59.39024	13.963964
## 1357	0.55440055	0.27720028	12.21413721	71.40010	4.250386
## 1358	0.89263051	0.22421525	0.72341146	55.02145	6.593634
## 1359	1.86750789	0.08832808	1.38801262	53.81222	5.794872
## 1360	0.30498136	0.40664182	0.71162318	59.37020	4.642857
## 1361	0.22866943	0.24296127	0.17150207	60.19490	4.111601
## 1362	0.00000000	0.06269592	0.00000000	57.44000	8.333333
## 1363	0.75624340	0.17587056	0.28139289	57.33006	5.961538
## 1364	0.00000000	0.31760436	2.90381125	57.49040	7.061503
## 1365	0.15450833	0.62906964	0.68425119	49.88060	6.210826
## 1366	0.52943008	0.15571473	0.31142946	44.93795	6.818182
## 1367	24.28018855	3.45322936	6.45311742	38.97061	6.377617
## 1486	11.57916158	4.15506258	1.69682585	50.37633	4.236256
## 1564	0.91285391	4.08768533	1.52655878	54.65082	5.072390
## 1744	5.04145306	2.74951939	2.40007209	53.42832	4.239802
## 1840	20.84468147	18.38740694	5.41566476	65.81954	5.338508
## 1924	3.54347316	2.51320172	3.71657786	58.84968	6.062214
## 2454	0.19477990	0.07791196	0.11686794	56.35408	7.093822
## 2455	3.10786948	2.03147793	0.42840691	46.24301	4.611181
## 2456	0.44897798	0.43322437	0.74435824	52.52364	5.656722
## 2457	1.23860847	0.58403634	0.15800017	51.24645	5.169122
## 2458	1.18790158	2.52207073	0.60701039	64.90515	5.473379
## 2459	0.36116273	0.49791367	0.12623164	54.46484	7.941812
## 2460	0.81454665	0.24683232	0.97087379	53.27641	8.914100
## 2461	1.16283390	0.75602779	0.44767248	60.30392	5.001246
## 2462	2.00561770	1.15310698	0.57655349	51.69798	6.780929
## 2463	0.25137111	0.42276051	0.04570384	55.39300	7.495430
## 2464	0.90489026	0.92414324	1.25144397	53.31825	2.465166
## 2465	0.61898212	2.73383769	3.71389271	53.63047	8.062361
## 2466	0.63919553	0.50279152	0.06661591	52.53935	6.153361
## 2467	5.21671857	4.56524081	2.65952759	54.81702	5.401030
## 2468	0.53721045	0.47313948	0.35978314	59.42298	6.386702
## 2469	0.57618788	0.24303659	0.17476789	56.35324	6.190278
## 2470	0.35839775	0.41461701	0.40056219	52.24349	3.506787
## 2471	0.40781078	0.38861968	0.19670873	55.92545	5.655207
## 2472	1.12955950	1.30757031	1.07453798	49.27269	5.846509
## 2473	1.11262048	0.57140393	0.28031136	54.15846	6.136245
## 2474	0.00000000	0.30196276	0.36906559	55.41502	8.227848
## 2475	12.16571379	6.79642938	2.53010580	43.89530	5.881342
## 2476	0.30299809	0.22326175	0.15415692	55.63023	3.366022
## 2477	0.55409740	0.54437640	0.42286381	58.75726	3.229308
## 2478	0.58230625	0.74420305	0.41518697	55.03236	5.876217
## 2479	0.48948274	0.39026326	0.27340477	55.68152	6.184256
## 2480	0.19586720	2.47772011	1.28293017	54.91502	7.996051
## 2481	0.43116916	0.33118790	0.13747422	54.82152	4.499230
## 2482	2.91678447	0.65262464	4.02648195	55.81103	5.730235

## 2483	0.60267857	0.20089286	0.08928571	51.67364	6.869221
## 2484	0.88095603	0.64348092	0.01532097	49.86311	3.367670
## 2485	0.62651289	0.81161897	0.12815036	55.75049	6.786942
## 2486	0.11162791	0.00000000	0.44651163	54.17731	7.011070
## 2487	0.58242593	0.91162320	0.00000000	51.32471	5.924855
## 2488	0.59190818	0.48363229	1.94535677	57.73595	7.140457
## 2489	0.37878788	0.15495868	1.13636364	54.22507	5.539070
## 2490	2.92618390	3.33086890	2.84057745	50.33139	5.799294
## 2491	0.37452145	0.62142818	1.46202075	53.36392	7.114267
## 2492	0.56404658	0.43668122	0.70960699	44.27786	8.348624
## 2493	0.26446631	0.19041574	0.32793822	54.68090	6.110202
## 2494	0.34889435	0.54054054	0.26535626	50.45861	7.700101
## 2495	0.43668122	0.41938692	0.23779670	57.37955	6.093113
## 2496	0.57344337	0.44170638	0.25572475	50.55744	6.653151
## 2497	0.54304957	0.50361022	0.26090650	55.11534	5.734926
## 2498	3.02342774	2.02411604	0.87949626	50.11345	6.166646
## 2499	0.35174112	0.43381405	0.62140931	55.47230	6.466069
## 2500	2.43607568	1.47796651	0.95810917	52.82287	5.329918
## 2501	3.98395352	5.75921059	7.20247153	53.92757	7.867172
## 2502	0.50806934	0.40346683	0.94142259	54.91202	6.289308
## 2503	5.10636295	5.86408132	0.79671364	53.62614	5.827671
## 2504	1.22585245	0.60510163	0.65204917	57.03768	7.511606
## 2505	1.65839830	0.83628632	0.29057406	45.21959	5.893846
## 2506	2.13224724	0.53734000	0.70504484	50.62935	4.865650
## 2507	1.25080180	0.40624332	1.26149241	52.63819	5.622271
## 2508	1.15700384	0.83367674	1.53421878	51.04928	7.867590
## 2509	0.54804531	0.35622945	0.44757033	56.62300	8.178625
## 2510	11.14671480	13.25095609	1.86988895	41.77379	6.136246
## 2511	0.66486087	0.27086924	0.91110564	54.60566	10.141207
## 2512	0.71233225	3.28181645	0.40068689	51.81544	5.744392
## 2513	0.75802518	0.58005405	2.76843979	54.09241	5.738255
## 2514	3.83595845	2.27784114	1.91566748	54.56684	4.613271
## 2515	0.53330545	0.93067029	0.35553696	56.98651	11.041340
## 2516	0.68523855	2.67050913	0.23695165	56.07787	6.330670
## 2517	1.54658714	1.03039310	0.21395869	45.37229	4.901104
## 2518	2.94126214	5.92039149	1.58022749	64.00598	6.064706
## 2519	1.97342950	1.20573887	0.25552815	62.64473	6.543533
## 2520	0.39944078	0.62579056	1.03854604	56.65115	5.582446
## 2521	3.74651856	1.89783414	1.19269963	51.70872	4.490661
## 2522	2.55181666	1.12805629	3.16786682	53.89251	6.135047
## 2523	0.77781189	1.86265480	0.90062430	53.92585	6.221294
## 2524	1.63779528	0.16797900	1.37532808	52.02080	9.562842
## 2525	0.28611134	0.42099240	1.92103327	55.95081	7.800673
## 2526	0.49577136	0.00000000	0.26246719	57.34908	4.967949
## 2527	0.16837379	0.35077873	0.28997708	57.73484	5.142332
## 2528	0.95210408	0.58143760	0.18169925	47.68200	5.574779
## 2529	2.78360243	0.76011743	0.79157056	56.88737	6.458604
## 2530	3.84075891	5.16925447	0.75809786	59.14859	5.104712
## 2531	0.59706893	1.14890537	2.66871721	50.04494	3.175334
## 2532	0.32178976	0.22984983	0.03064664	51.90922	7.810107
## 2533	1.43127721	2.55403901	0.44519946	49.48605	4.802427
## 2534	1.01327987	1.01949631	1.01871926	62.21363	6.296560
## 2535	0.45580658	0.59453032	1.39714625	56.38704	7.744283
## 3019	1.70476263	0.51219326	0.59628469	48.20957	8.708822

## 3020	0.11398176	0.00000000	0.00000000	55.73171	8.419023
## 3021	4.09427043	0.33668009	0.40281368	47.10757	6.012885
## 3022	0.33542977	0.46121593	0.25157233	57.39525	8.417508
## 3023	1.70079200	0.17883864	3.93080039	50.09729	7.120636
## 3024	3.52416959	0.87091547	0.14852822	52.63251	12.159893
## 3025	1.75101215	0.56680162	0.16194332	50.87762	7.974482
## 3026	2.04260727	0.93959935	0.40095624	57.05104	5.706173
## 3027	0.88170463	0.00000000	0.44085231	51.44144	5.031447
## 3028	0.56818182	0.00000000	0.00000000	53.43610	2.613240
## 3029	0.00000000	1.15585384	3.24384788	52.77338	10.775862
## 3030	1.02564103	0.74592075	1.72494173	50.48544	6.746988
## 3031	0.60562878	0.54625341	0.20187626	59.69796	6.744329
## 3032	0.42831138	0.81379163	1.76678445	45.56004	7.082002
## 3033	0.78032632	0.63844881	0.10640813	55.11348	7.435159
## 3034	0.00000000	0.54726368	0.99502488	57.78061	4.664723
## 3035	3.16804789	1.61035505	1.89306799	50.73057	7.404551
## 3036	0.41308090	0.00000000	0.41308090	42.70197	7.127430
## 3037	0.51752408	0.29206805	0.79934413	55.21606	6.565531
## 3038	3.69003690	0.24176104	0.08906986	55.93438	4.685714
## 3039	3.99940853	4.40735168	0.60713074	41.84409	4.146877
## 3040	0.90270812	0.30090271	3.44366433	55.96546	4.210526
## 3041	0.78270593	0.40998882	0.44726053	52.14968	3.050109
## 3042	1.14510399	1.38654158	1.01403787	43.27678	3.977583
## 3043	3.83775351	0.32761311	1.70046802	51.06383	7.773512
## 3044	2.32677134	4.04491961	14.13028842	52.00794	8.186470
## 3045	2.31318776	1.31647174	5.68070519	55.15395	7.809192
## 3046	1.17656195	0.24463169	2.13179047	58.48423	7.582938
## 3047	16.59009976	3.17775301	1.35645744	56.04024	8.981723
##	deathRate	PctDoubleCoverage	PctNoCoverage	EmpSponsoredPct	
## 112	211.0	0.0	0.0	55.39683	
## 113	167.7	0.0	9.8	72.17806	
## 114	145.2	6.5	0.0	56.01132	
## 115	143.2	0.0	8.5	83.42466	
## 116	262.1	0.0	6.4	61.76991	
## 117	178.7	0.0	5.0	68.84817	
## 118	191.5	0.0	10.0	77.66571	
## 119	123.8	0.0	6.5	77.71346	
## 120	127.4	0.0	4.4	75.66303	
## 121	195.7	0.0	1.4	63.17512	
## 122	254.3	1.3	0.0	62.65060	
## 123	213.5	1.3	0.0	56.87382	
## 124	121.8	0.0	5.4	63.60601	
## 125	132.6	2.6	0.0	59.27007	
## 126	166.0	0.0	6.5	69.87768	
## 127	188.4	0.0	0.5	71.59763	
## 128	193.5	0.0	3.1	67.69984	
## 149	160.3	0.5	0.0	76.47059	
## 368	164.1	17.8	0.0	49.01961	
## 369	167.9	5.0	0.0	76.41866	
## 370	189.3	5.1	0.0	61.46927	
## 371	162.6	0.0	0.9	67.57679	
## 372	155.3	3.7	0.0	71.74825	
## 568	182.7	0.0	2.1	67.91277	
## 1014	153.4	5.8	0.0	68.54220	

## 1026	142.3	0.0	4.9	62.81646
## 1295	136.2	7.0	0.0	46.98795
## 1296	134.0	15.8	0.0	48.31169
## 1297	197.8	0.0	12.1	65.16854
## 1298	150.7	0.0	1.8	63.60053
## 1299	156.9	10.7	0.0	52.13675
## 1300	155.4	7.4	0.0	58.35735
## 1301	164.2	0.0	10.0	60.61121
## 1302	194.7	4.5	0.0	62.06416
## 1303	162.2	5.0	0.0	68.17577
## 1304	122.6	0.0	10.1	66.34615
## 1305	127.6	7.4	0.0	59.34343
## 1306	178.7	8.7	0.0	67.69627
## 1307	173.0	6.7	0.0	69.84536
## 1308	165.8	16.3	0.0	42.64706
## 1309	146.8	2.9	0.0	75.00000
## 1310	209.6	0.0	1.1	70.70968
## 1311	129.4	3.5	0.0	60.55944
## 1312	141.4	7.6	0.0	59.78112
## 1313	148.6	7.4	0.0	62.14834
## 1314	166.5	6.9	0.0	53.24190
## 1315	164.5	3.8	0.0	57.62712
## 1316	219.4	7.3	0.0	59.73333
## 1317	129.1	12.6	0.0	56.63717
## 1318	182.9	0.0	3.2	68.94502
## 1319	161.2	8.5	0.0	69.39281
## 1320	143.8	8.3	0.0	65.55407
## 1321	150.8	10.3	0.0	61.90476
## 1322	176.2	3.3	0.0	65.44021
## 1323	150.2	5.6	0.0	69.05055
## 1324	151.3	7.1	0.0	59.49686
## 1325	208.4	1.0	0.0	66.66667
## 1326	193.4	10.5	0.0	55.86298
## 1327	165.4	1.0	0.0	69.95769
## 1328	147.7	8.4	0.0	62.43781
## 1329	183.5	2.8	0.0	68.14159
## 1330	128.7	12.7	0.0	55.18135
## 1331	109.9	5.7	0.0	63.04945
## 1332	181.3	7.1	0.0	62.44784
## 1333	166.2	9.4	0.0	52.14477
## 1334	151.2	5.7	0.0	65.29563
## 1335	182.5	5.6	0.0	57.82967
## 1336	171.1	9.2	0.0	58.70712
## 1337	153.9	4.2	0.0	64.80100
## 1338	177.9	3.9	0.0	65.45961
## 1339	177.8	12.7	0.0	52.57732
## 1340	172.6	2.2	0.0	66.42755
## 1341	154.1	11.6	0.0	56.22490
## 1342	173.6	0.6	0.0	67.61769
## 1343	154.3	1.8	0.0	59.25059
## 1344	164.7	7.0	0.0	57.06294
## 1345	114.3	8.7	0.0	60.37234
## 1346	151.9	9.8	0.0	48.36879
## 1347	175.6	0.0	0.0	66.71368

## 1348	160.5	8.8	0.0	60.40428
## 1349	171.5	0.0	2.8	72.68786
## 1350	164.4	0.0	12.5	79.67480
## 1351	171.6	3.6	0.0	67.64296
## 1352	143.1	11.0	0.0	43.05177
## 1353	161.6	2.0	0.0	60.49005
## 1354	157.6	15.1	0.0	41.05263
## 1355	126.0	1.5	0.0	60.12365
## 1356	127.9	0.4	0.0	67.88512
## 1357	148.2	0.0	5.1	64.70588
## 1358	170.1	3.5	0.0	69.46355
## 1359	154.5	5.7	0.0	59.62963
## 1360	177.9	10.1	0.0	56.66267
## 1361	150.3	6.6	0.0	68.23239
## 1362	175.5	2.1	0.0	53.38645
## 1363	172.4	8.3	0.0	60.08065
## 1364	177.6	4.9	0.0	45.74616
## 1365	171.1	0.1	0.0	62.06323
## 1366	293.9	1.6	0.0	56.73534
## 1367	221.8	0.0	13.3	71.59533
## 1486	160.8	0.0	5.9	70.35398
## 1564	160.2	3.6	0.0	68.79150
## 1744	172.7	7.2	0.0	70.29831
## 1840	133.4	0.0	8.7	81.08108
## 1924	164.9	0.0	8.3	75.73099
## 2454	178.0	14.8	0.0	46.38633
## 2455	151.9	6.0	0.0	70.81761
## 2456	161.5	12.1	0.0	63.92638
## 2457	187.8	7.2	0.0	72.64022
## 2458	148.8	4.5	0.0	78.40000
## 2459	168.1	11.2	0.0	51.70178
## 2460	167.2	8.0	0.0	58.68613
## 2461	149.0	7.1	0.0	73.53308
## 2462	161.0	5.3	0.0	73.35025
## 2463	184.9	2.8	0.0	59.89583
## 2464	128.7	12.5	0.0	58.05142
## 2465	153.8	14.1	0.0	51.69972
## 2466	176.7	11.2	0.0	60.48502
## 2467	154.4	4.9	0.0	77.49392
## 2468	153.6	6.5	0.0	73.47188
## 2469	160.9	13.1	0.0	59.43517
## 2470	158.9	11.7	0.0	58.39112
## 2471	152.7	8.0	0.0	62.16578
## 2472	164.1	9.3	0.0	62.01991
## 2473	161.6	8.5	0.0	66.83938
## 2474	159.9	14.7	0.0	52.51701
## 2475	156.8	2.7	0.0	74.40318
## 2476	141.8	10.9	0.0	65.97132
## 2477	156.0	12.8	0.0	54.05007
## 2478	177.3	6.1	0.0	72.93333
## 2479	177.7	10.2	0.0	62.06897
## 2480	130.4	12.6	0.0	59.32836
## 2481	162.2	9.1	0.0	61.76471
## 2482	147.5	6.8	0.0	63.08345

## 2483	168.3	10.3	0.0	56.23306
## 2484	173.3	9.8	0.0	59.87159
## 2485	145.6	17.0	0.0	48.98219
## 2486	164.5	14.4	0.0	61.61335
## 2487	172.7	9.5	0.0	59.22190
## 2488	166.9	8.3	0.0	71.28335
## 2489	146.8	12.3	0.0	57.25191
## 2490	163.0	6.2	0.0	67.01434
## 2491	175.4	9.4	0.0	67.91721
## 2492	215.7	0.0	3.8	57.82881
## 2493	148.2	8.9	0.0	60.41667
## 2494	162.5	11.3	0.0	59.03448
## 2495	151.2	8.3	0.0	65.78600
## 2496	190.0	8.1	0.0	64.13199
## 2497	169.5	8.9	0.0	62.41234
## 2498	155.6	8.4	0.0	66.75749
## 2499	156.4	12.5	0.0	53.28084
## 2500	133.3	8.7	0.0	71.25155
## 2501	138.1	0.6	0.0	63.15007
## 2502	169.9	10.8	0.0	55.76102
## 2503	142.2	6.8	0.0	75.74627
## 2504	156.9	13.0	0.0	59.03614
## 2505	174.2	9.4	0.0	69.00662
## 2506	161.7	8.8	0.0	61.54971
## 2507	173.0	6.9	0.0	53.54559
## 2508	184.4	7.0	0.0	64.00000
## 2509	154.5	14.4	0.0	56.25000
## 2510	166.5	2.4	0.0	73.54651
## 2511	157.6	8.9	0.0	67.21536
## 2512	180.7	8.1	0.0	55.02793
## 2513	176.5	11.2	0.0	56.56168
## 2514	168.4	4.3	0.0	70.02551
## 2515	147.0	8.0	0.0	60.16260
## 2516	158.2	10.0	0.0	68.53234
## 2517	184.8	9.3	0.0	66.35135
## 2518	153.9	2.3	0.0	79.53216
## 2519	170.5	4.6	0.0	78.37181
## 2520	181.7	8.8	0.0	65.66580
## 2521	153.7	6.5	0.0	70.21823
## 2522	145.9	8.2	0.0	71.74770
## 2523	146.2	12.0	0.0	62.11401
## 2524	162.3	10.6	0.0	58.88889
## 2525	147.2	3.4	0.0	62.17852
## 2526	132.3	17.5	0.0	50.65963
## 2527	160.2	11.9	0.0	62.98201
## 2528	187.9	10.2	0.0	55.36000
## 2529	163.0	8.6	0.0	67.53247
## 2530	152.1	6.9	0.0	75.96266
## 2531	159.7	3.9	0.0	60.08646
## 2532	122.5	9.5	0.0	62.19975
## 2533	166.8	4.9	0.0	65.85678
## 2534	165.2	4.6	0.0	76.38376
## 2535	151.9	9.5	0.0	56.37584
## 3019	196.4	3.7	0.0	64.69697

## 3020	174.8	5.3	0.0	50.07386
## 3021	180.2	1.3	0.0	66.37807
## 3022	197.8	8.1	0.0	54.60705
## 3023	178.3	0.2	0.0	64.04834
## 3024	166.6	2.8	0.0	66.56535
## 3025	184.2	4.6	0.0	61.46927
## 3026	175.0	3.1	0.0	71.93211
## 3027	177.8	8.2	0.0	60.76818
## 3028	194.2	5.7	0.0	54.72000
## 3029	150.0	13.2	0.0	49.30991
## 3030	153.7	4.8	0.0	69.58831
## 3031	169.2	14.0	0.0	53.99002
## 3032	178.6	6.0	0.0	58.24022
## 3033	176.6	5.1	0.0	68.76712
## 3034	209.6	6.8	0.0	58.84718
## 3035	184.4	3.4	0.0	66.95652
## 3036	214.7	7.8	0.0	52.50737
## 3037	181.2	8.5	0.0	57.16216
## 3038	160.3	0.0	0.7	65.47278
## 3039	153.7	0.0	1.1	69.73518
## 3040	208.3	0.0	1.3	57.85609
## 3041	215.9	11.5	0.0	54.84848
## 3042	146.4	4.3	0.0	66.12500
## 3043	149.6	10.0	0.0	56.96041
## 3044	150.1	0.0	6.7	75.34884
## 3045	153.9	0.0	11.4	77.09677
## 3046	175.0	5.4	0.0	65.34914
## 3047	213.6	1.2	0.0	30.09198

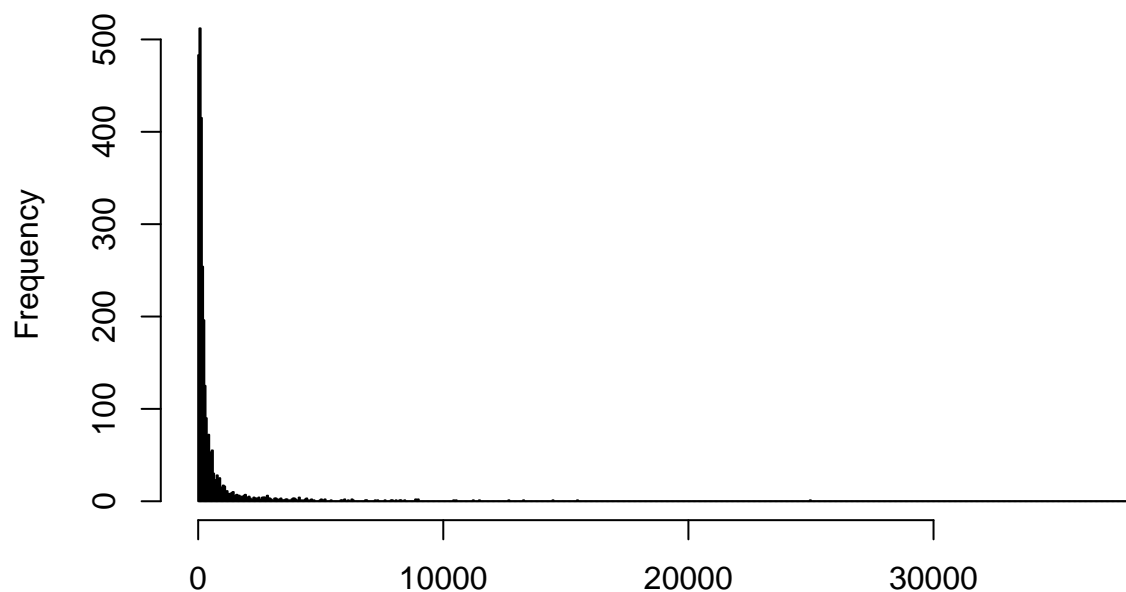
We notice that there are a lot of observations that have the exact same value of avgAnnCount - 1962.667684 - and this is the only not integer value in the whole set of avgAnnCount values. For that matter, before any further analysis, we will transform all of these values in NAs.

```
cancer.df$avgAnnCount[cancer.df$avgAnnCount==1962.667684] = NA
```

Now we can go back to analyzing the avgAnnCount:

```
hist(cancer.df$avgAnnCount, breaks="fd", xlab = "Mean Number of Incidences per County (2009-2013)", ylab = "Frequency")
```

Corrected Histogram of Mean Cancer Incidences



Mean Number of Incidences per County (2009–2013)

That extremely right-skewed distribution is an indicative that we could use a `log()` transformation in this variable. Let's see how the variable is before the transformation.

```
summary(cancer.df$avgAnnCount)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   NA's  
##         6      71     153     508    396   38150    206
```

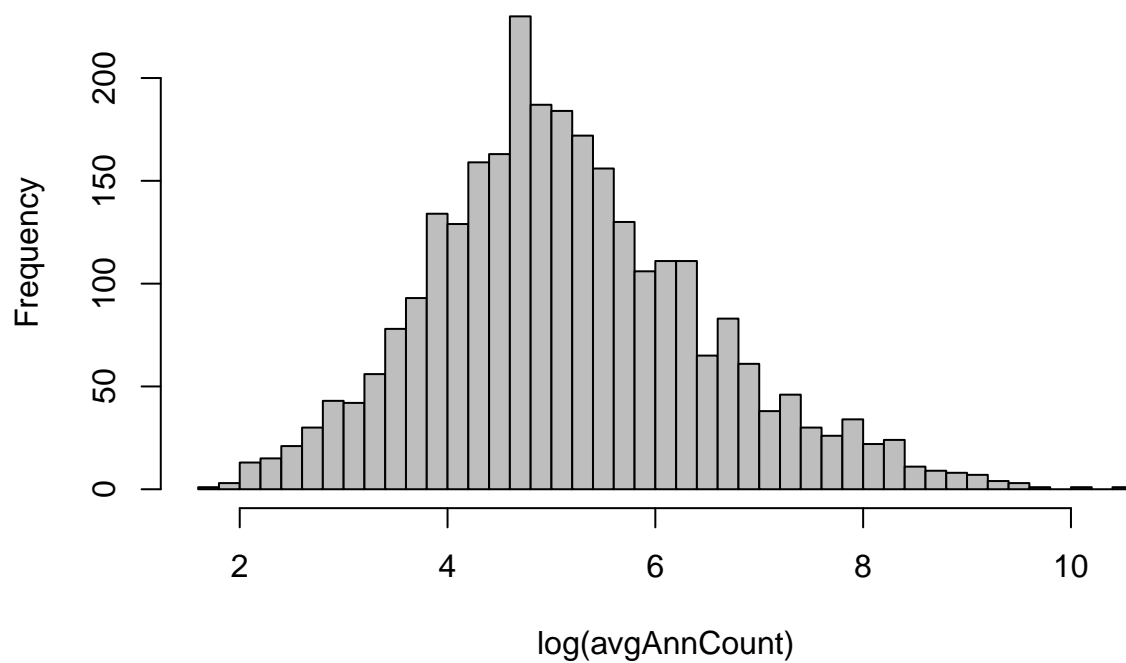
Now to the analysis of the `log(avgAnnCount)`

```
summary(log(cancer.df$avgAnnCount))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   NA's  
##   1.792   4.263   5.030   5.158   5.981  10.549    206
```

```
hist(log(cancer.df$avgAnnCount), breaks="fd", xlab = "log(avgAnnCount)", ylab="Frequency", main = "Histogram of log(avgAnnCount)")
```

Histogram of $\log(\text{avgAnnCount})$

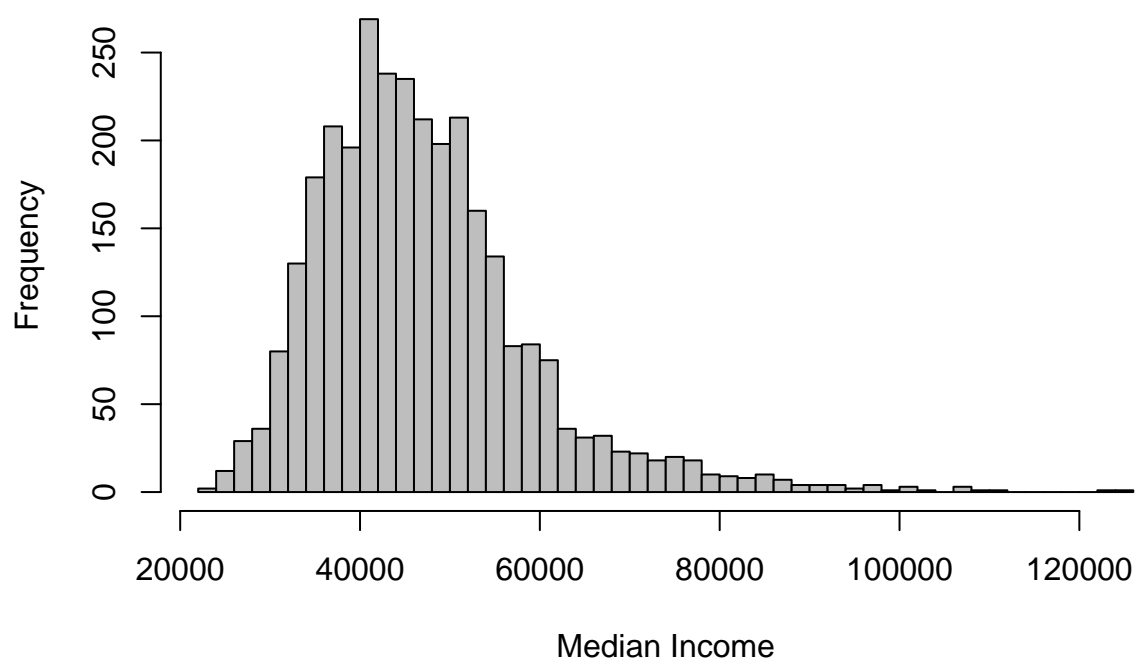


The distribution of $\log(\text{avgAnnCount})$ is fairly close to normal.

medIncome

```
hist(cancer.df$medIncome, breaks="fd", xlab = "Median Income", ylab="Frequency", main = "Histogram of M
```

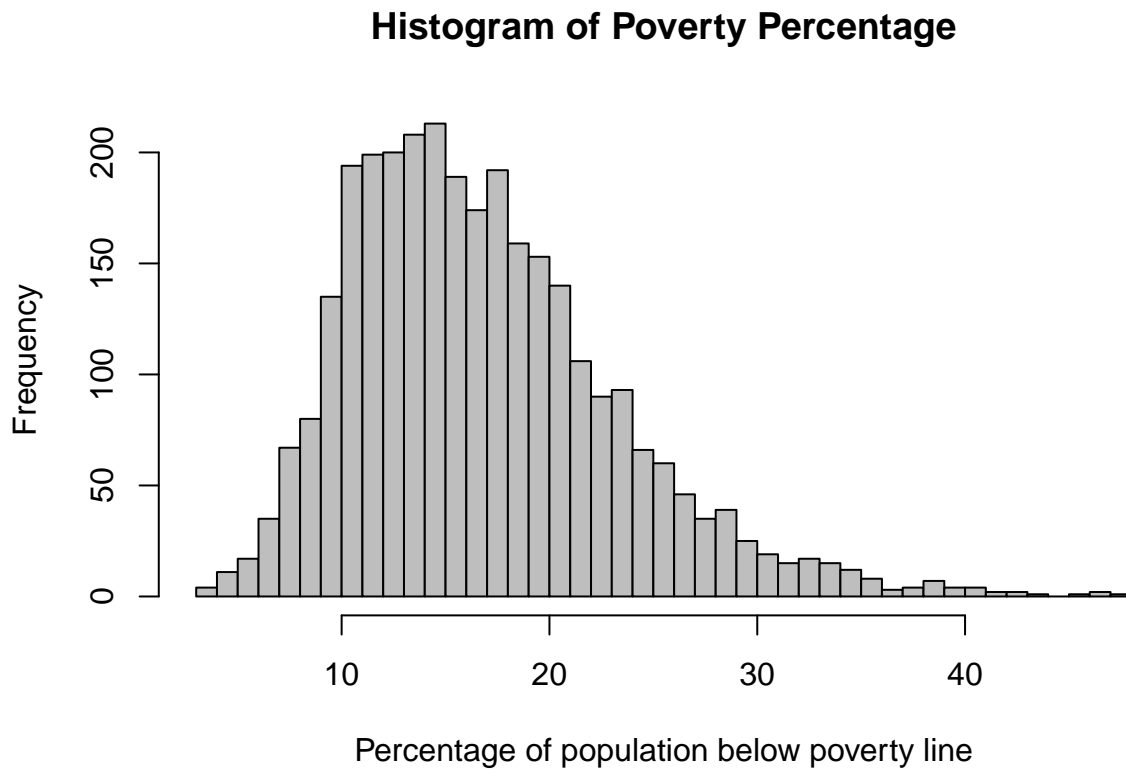
Histogram of Median Income



The distribution of Median Income is fairly normal with a slight right skew.

povertyPercent

```
hist(cancer.df$povertyPercent, breaks="fd", xlab = "Percentage of population below poverty line", ylab=
```



The distribution of Poverty Percent is fairly normal with a slight right skew.

Bivariate Analysis of Secondary Variables

After taking a first look into other variables, and performing necessary corrections, we must now understand how they relate to the primary key variables, in order to comprehend what else might be driving the relationship previously found between different types of health insurance coverage and cancer death rates.

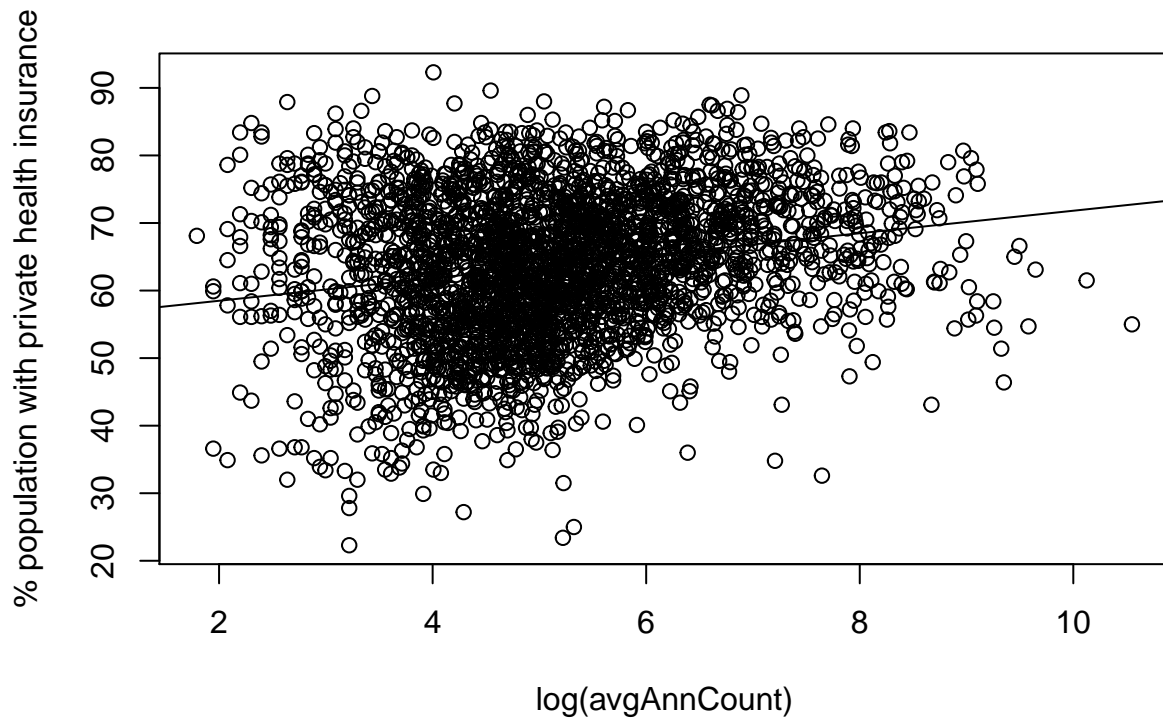
For each secondary variable introduced, we will analyze their relationship with the primary variables and with the output variable itself.

log(avgAnnCount)

Private Insurance Coverage

```
plot(log(cancer.df$avgAnnCount),cancer.df$PctPrivateCoverage, ylab = "% population with private health insurance",
abline(lm(cancer.df$PctPrivateCoverage[!is.na(cancer.df$avgAnnCount)] ~ log(cancer.df$avgAnnCount[!is.na(cancer.df$avgAnnCount)])))
```

Private Coverage vs log(avgAnnCount)



```
cor(log(cancer.df$avgAnnCount),cancer.df$PctPrivateCoverage, use = "complete.obs")
```

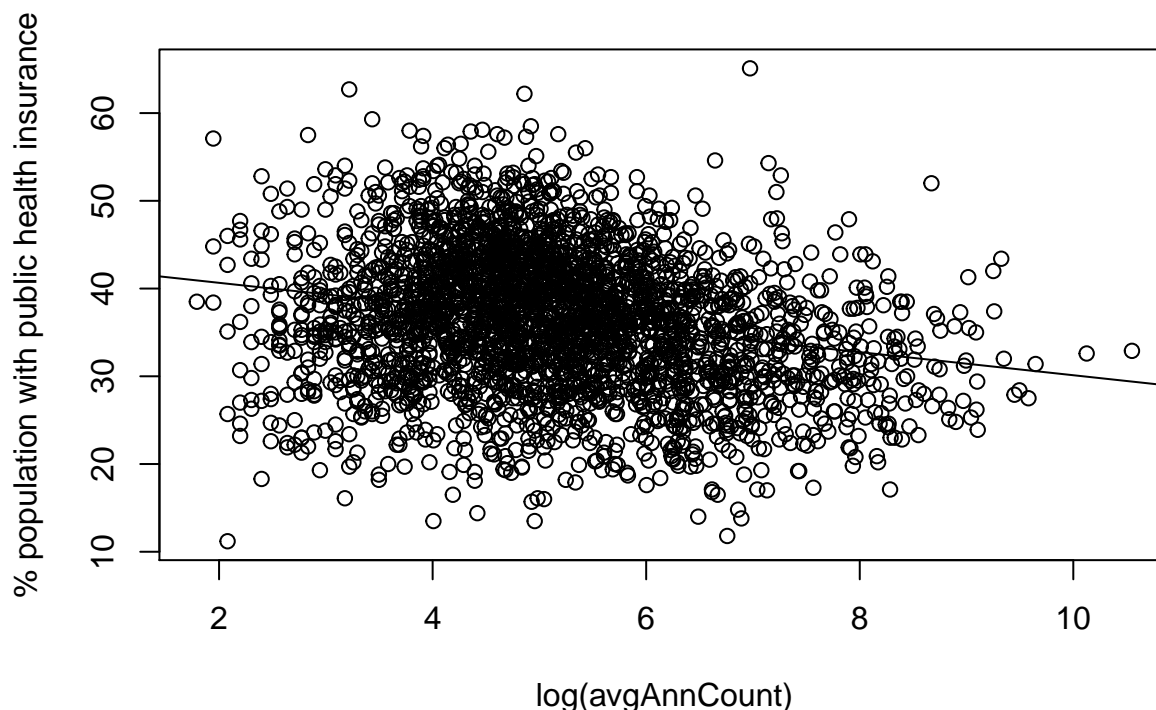
```
## [1] 0.2103135
```

There is a small positive correlation between the $\log(\text{avgAnnCount})$ and the percentage of population with private health insurance coverage.

Public Insurance Coverage

```
plot(log(cancer.df$avgAnnCount),cancer.df$PctPublicCoverage, ylab = "% population with public health insurance",  
abline(lm(cancer.df$PctPublicCoverage[!is.na(cancer.df$avgAnnCount)] ~ log(cancer.df$avgAnnCount[!is.na(cancer.df$avgAnnCount)])))
```

Public Coverage vs log(avgAnnCount)



```
cor(log(cancer.df$avgAnnCount),cancer.df$PctPublicCoverage, use = "complete.obs")
```

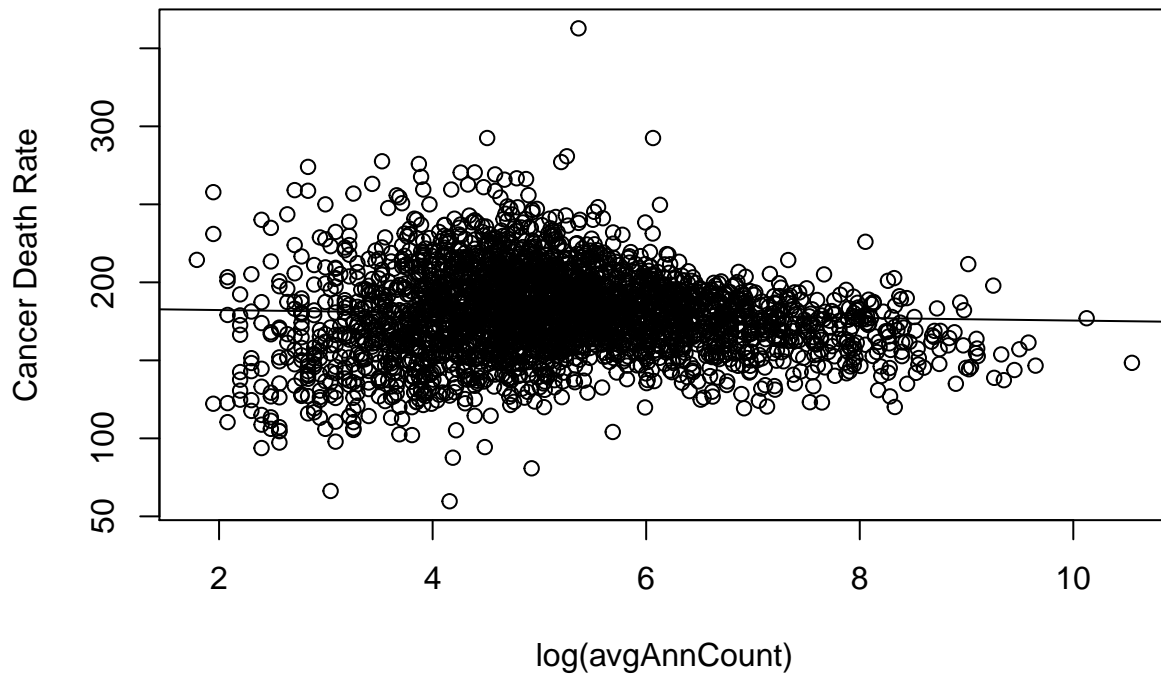
```
## [1] -0.2241446
```

There is a small negative correlation between the $\log(\text{avgAnnCount})$ and the percentage of population with public health insurance coverage, in a very close magnitude to the positive correlation encountered with the private health insurance coverage. We have seen before that, in terms of death rate, these two types of health insurance coverage have opposite behaviors. That being said, even if it is a small correlation, the fact that it presents itself in opposite ways and in similar magnitude to public and private health insurance coverage, just like death rates, indicates that we should dig deeper to check if the $\log(\text{avgAnnCount})$ has a stronger positive correlation with death rate.

Death Rate

```
plot(log(cancer.df$avgAnnCount),cancer.df$deathRate, ylab = "Cancer Death Rate", xlab="log(avgAnnCount)")
abline(lm(cancer.df$deathRate[!is.na(cancer.df$avgAnnCount)] ~ log(cancer.df$avgAnnCount[!is.na(cancer.df$avgAnnCount)]))
```


Death Rate vs log(avgAnnCount)



```
cor(log(cancer.df$avgAnnCount), cancer.df$deathRate, use = "complete.obs")
```

```
## [1] -0.04059046
```

At first sight, by just analyzing these charts and the correlation, it seems as we don't have much of a relation between these two variables. However, since they presented the same opposite behavior with public and private health insurance coverage, we want to take a deeper look into what might be going on.

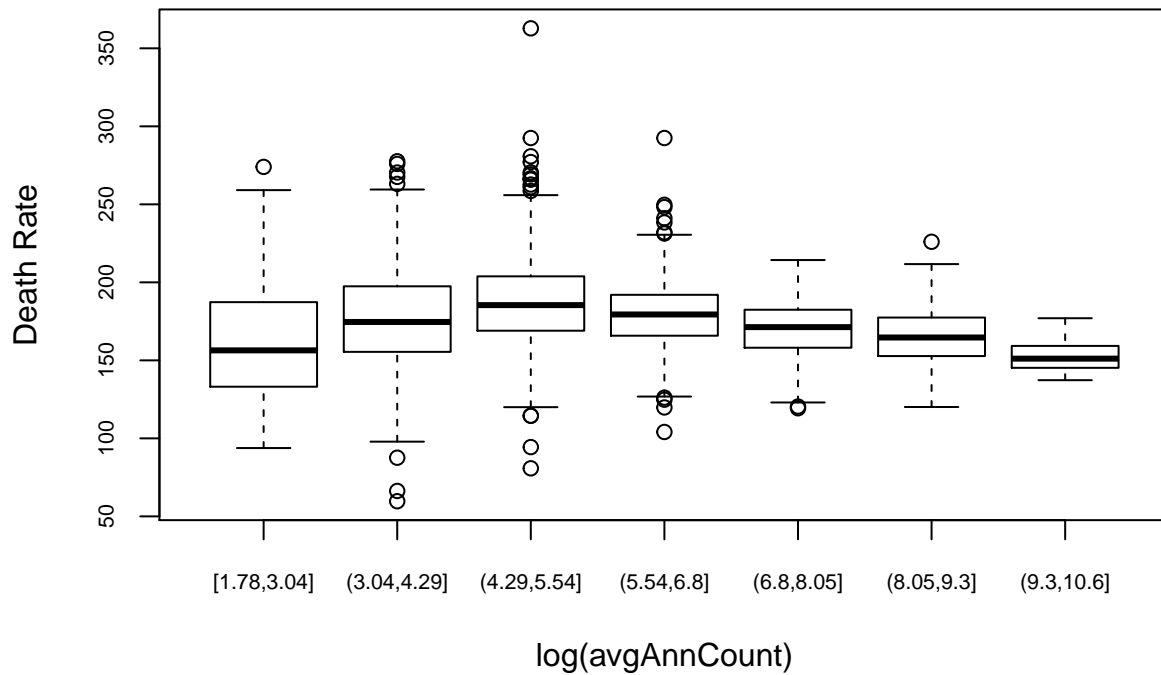
```
levels(cut(log(cancer.df$avgAnnCount), 7, include.lowest=TRUE))
```

```
## [1] "[1.78,3.04]" "(3.04,4.29]" "(4.29,5.54]" "(5.54,6.8]" "(6.8,8.05]"
```

```
## [6] "(8.05,9.3]" "(9.3,10.6]"
```

```
boxplot(deathRate ~ cut(log(avgAnnCount), 7, include.lowest=TRUE),
        data = cancer.df,
        cex.axis = .7,
        main = "Death Rate for different levels of incidence rate",
        xlab = "log(avgAnnCount)", ylab = "Death Rate")
```

Death Rate for different levels of incidence rate



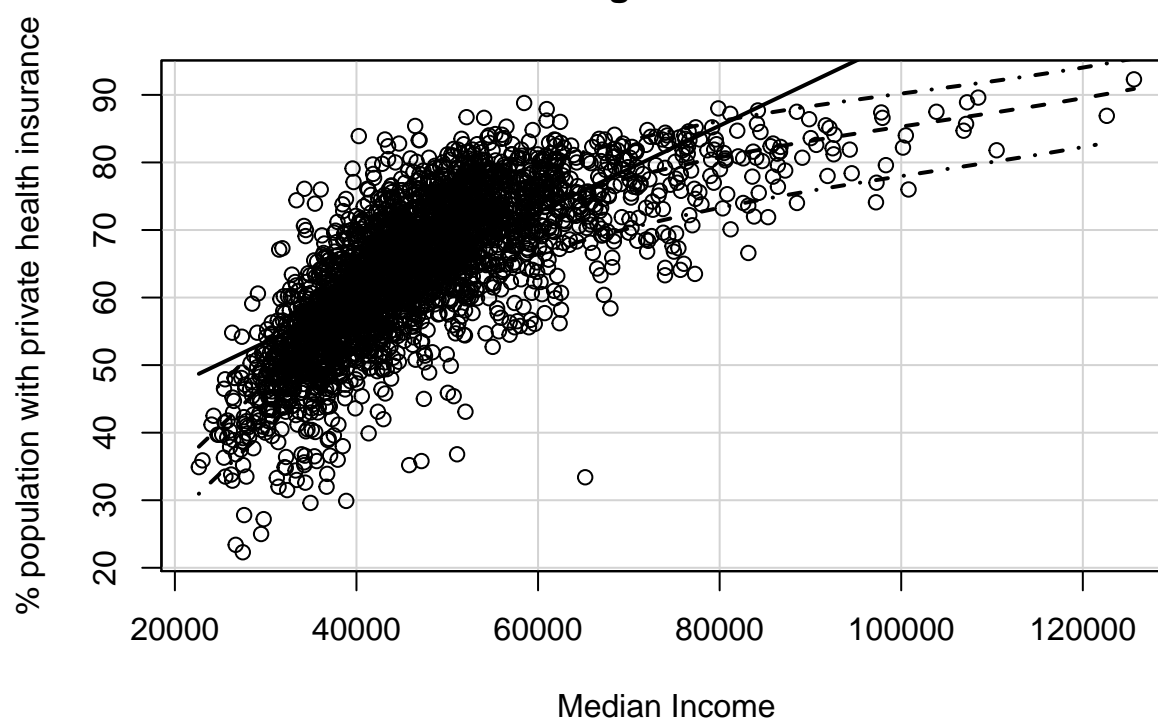
With this visualization it seems as the incidence has a positive correlation with the death rate up to a certain point. But passed that threshold (4.29, 5.54], the correlation becomes negative. One possible interpretation we had is that after a certain number of cancer reported cases, there is a more pressing need to invest in that disease, increasing the survival chances for those with it and, consequently, decreasing the death rates.

medIncome

Private Insurance Coverage

```
scatterplot(cancer.df$medIncome,cancer.df$PctPrivateCoverage, ylab = "% population with private health insurance")
```

Private Coverage vs Median Income



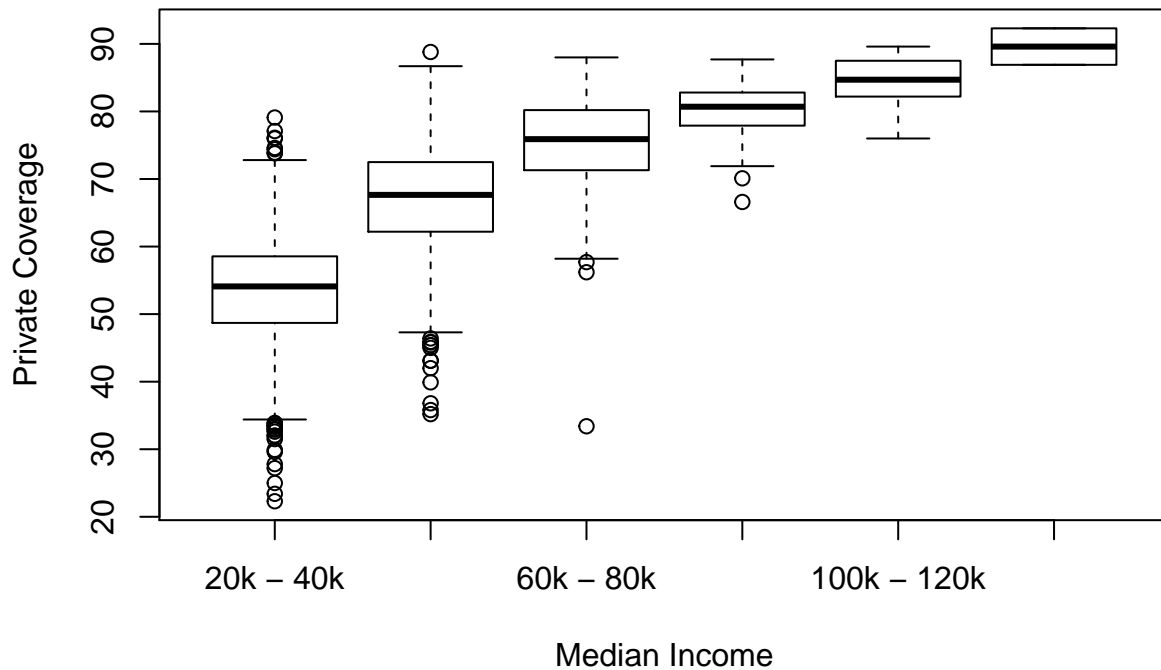
```
cor(cancer.df$medIncome,cancer.df$PctPrivateCoverage)
```

```
## [1] 0.7241748
```

By the chart presented and the strong positive correlation we attest something probably intuitively known: populations with higher income tend to have more private health insurance coverage. We can take a deeper look by analyzing boxplots for different levels of median income:

```
boxplot(cancer.df$PctPrivateCoverage ~ cut(cancer.df$medIncome, right=FALSE,seq(20000,140000,20000),labels=
  main = "Private Coverage for different levels of income",
  xlab = "Median Income", ylab = "Private Coverage")
```

Private Coverage for different levels of income

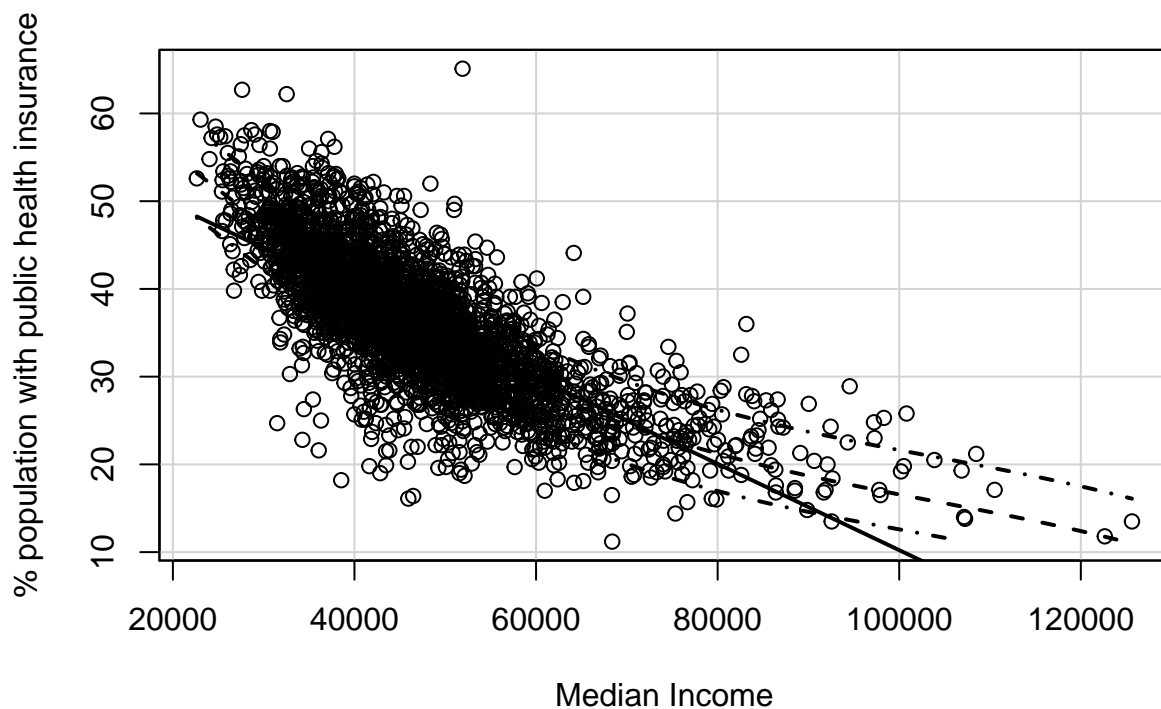


It confirms what we previously stated. There is a clear correlation between wealth and private health insurance coverage.

Public Insurance Coverage

```
scatterplot(cancer.df$medIncome, cancer.df$PctPublicCoverage, ylab = "% population with public health insurance")
```

Public Coverage vs Median Income

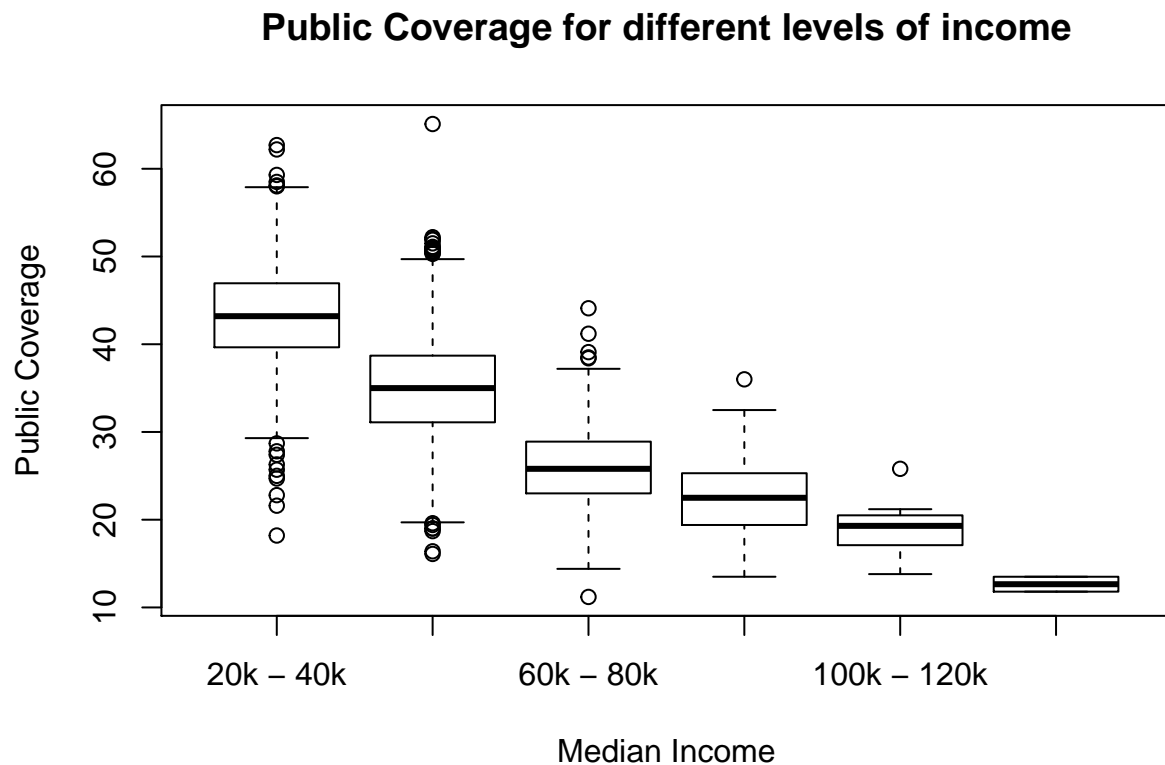


```
cor(cancer.df$medIncome,cancer.df$PctPublicCoverage)
```

```
## [1] -0.7548218
```

Similarly to what we have seen with the Private Health Insurance Coverage, there is a clear negative correlation between Public Health Insurance Coverage and the median income. That can be interpreted that populations with lower income tend to be more dependent on Public Health Insurance, probably because the Private option is not affordable. We can take a deeper look by analyzing boxplots for different levels of median income:

```
boxplot(cancer.df$PctPublicCoverage ~ cut(cancer.df$medIncome, right=FALSE,seq(20000,140000,20000),label=
  main = "Public Coverage for different levels of income",
  xlab = "Median Income", ylab = "Public Coverage")
```

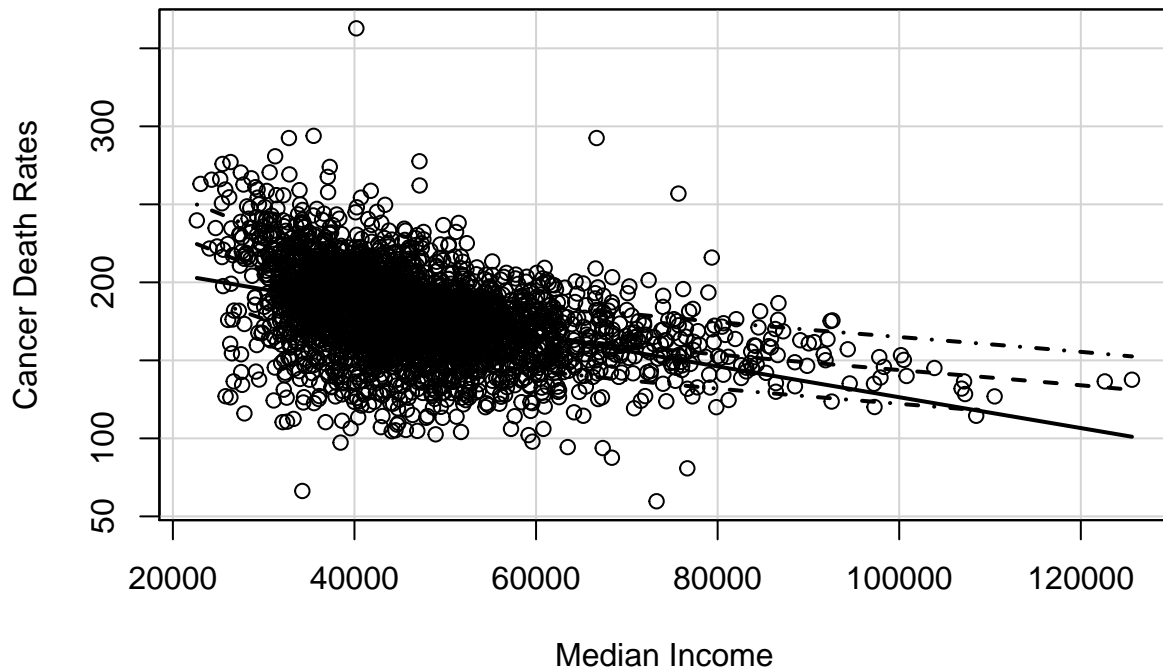


The boxplot only confirms what we have noticed before. For higher levels of income, the public health insurance coverage is lower. That means this variable presents also presents the opposite behavior as death rates. The higher the median income, higher the private coverage and lower the public coverage. So we should probably check if there is a direct relation between the median income and death rates.

Death Rates

```
scatterplot(cancer.df$medIncome,cancer.df$deathRate, ylab = "Cancer Death Rates", xlab="Median Income",
```

Median Income vs Cancer Death Rates



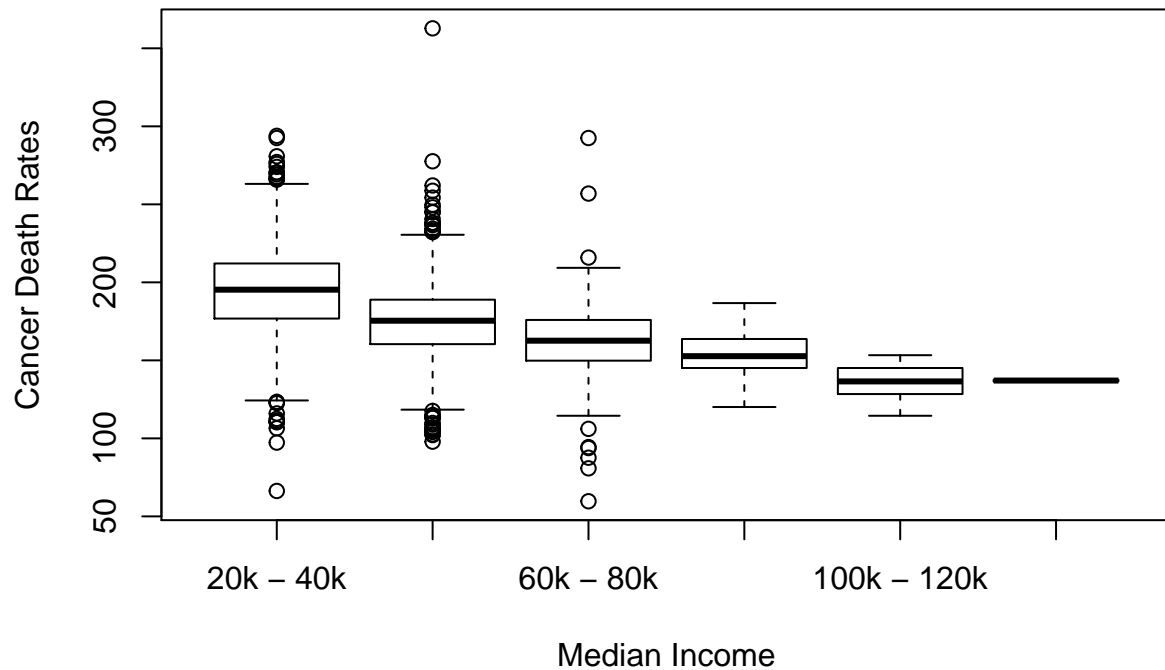
```
cor(cancer.df$medIncome,cancer.df$deathRate)
```

```
## [1] -0.4286149
```

We see that there is a stronger negative correlation between median income and death rates than private health insurance coverage and death rates, which may lead us to the hypothesis that actually socioeconomic factors have more to do with death rates than the percent coverage by type of health insurance itself. Taking a deeper look by analyzing the boxplot by levels of median income may provide us with better insights

```
boxplot(cancer.df$deathRate ~ cut(cancer.df$medIncome, right=FALSE,seq(20000,140000,20000),labels = c("20000-40000", "40000-60000", "60000-80000", "80000-100000", "100000-120000"),
  main = "Cancer Death Rates for different levels of income",
  xlab = "Median Income", ylab = "Cancer Death Rates")
```

Cancer Death Rates for different levels of income



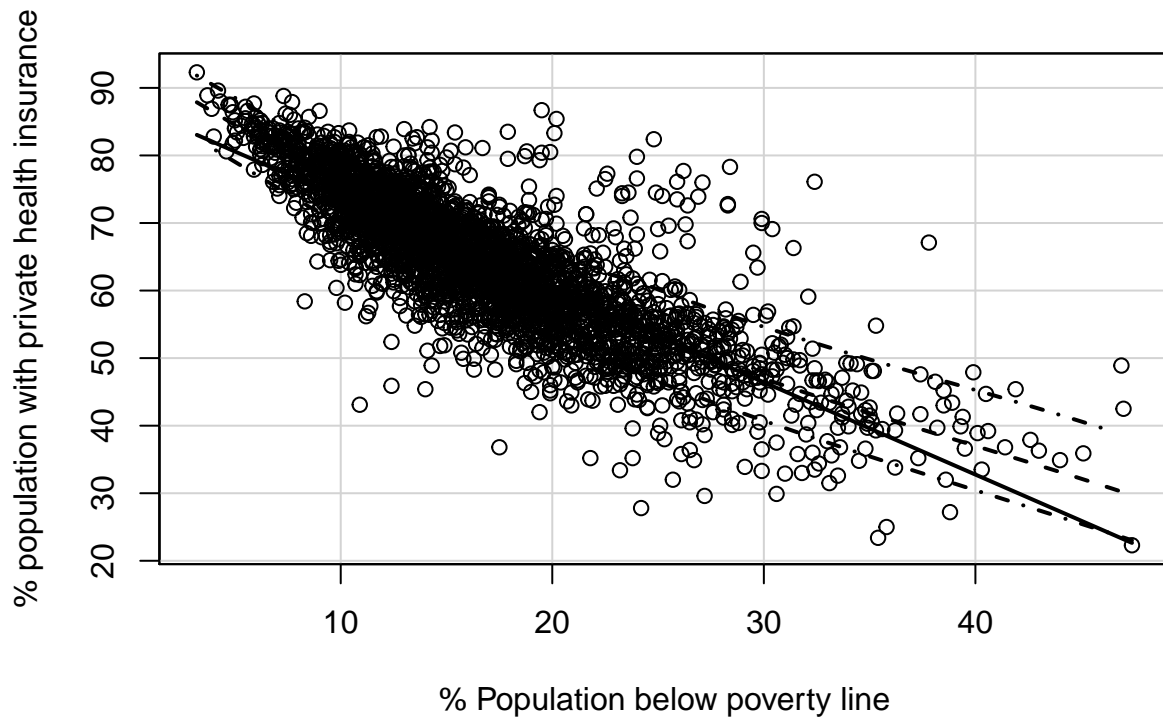
It seems to confirm our hypothesis, however, taking a deeper look into another socioeconomic variable might strengthen our hypothesis.

povertyPercent

Private Insurance Coverage

```
scatterplot(cancer.df$povertyPercent, cancer.df$PctPrivateCoverage, ylab = "% population with private health insurance")
```

Private Coverage vs Poverty Percent



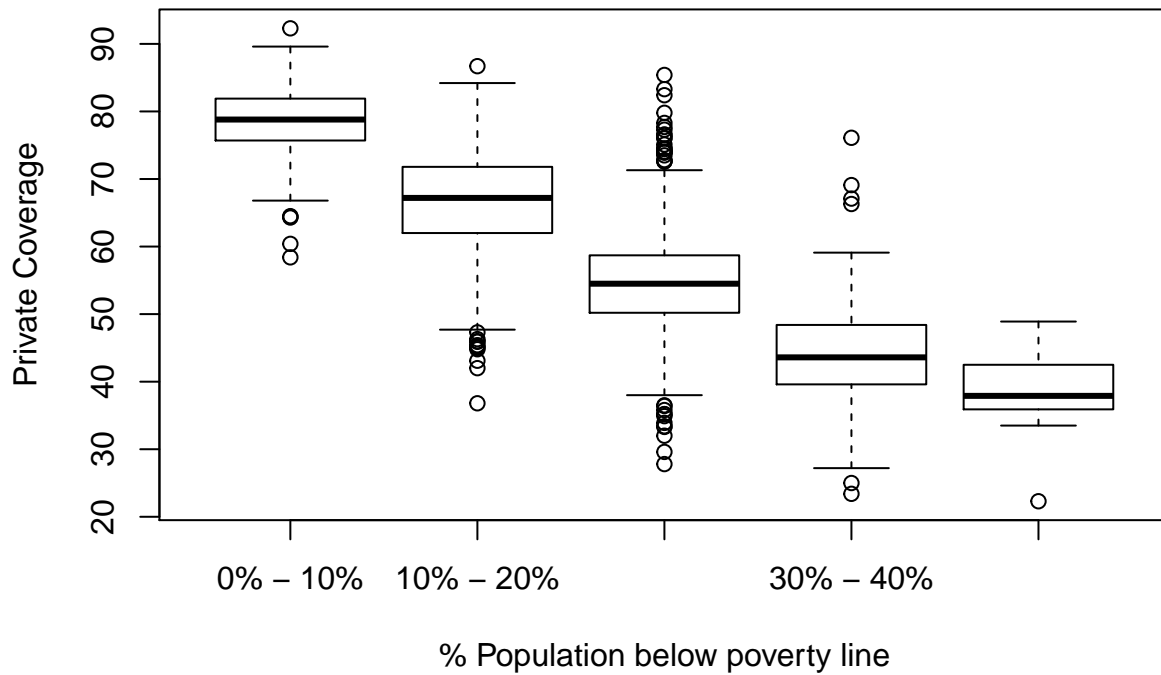
```
cor(cancer.df$povertyPercent,cancer.df$PctPrivateCoverage)
```

```
## [1] -0.8225343
```

The strongest relation we have encountered so far, we see that populations with higher percentage below poverty line tend to have less private health insurance coverage, the opposite behavior to the median income variable. Taking a look into boxplots provides us with a indicative of validity of such hypothesis:

```
boxplot(cancer.df$PctPrivateCoverage ~ cut(cancer.df$povertyPercent, right=FALSE,seq(0,50,10),labels = c(
  main = "Private Coverage for different levels of poverty percent",
  xlab = "% Population below poverty line", ylab = "Private Coverage")
```


Private Coverage for different levels of poverty percent

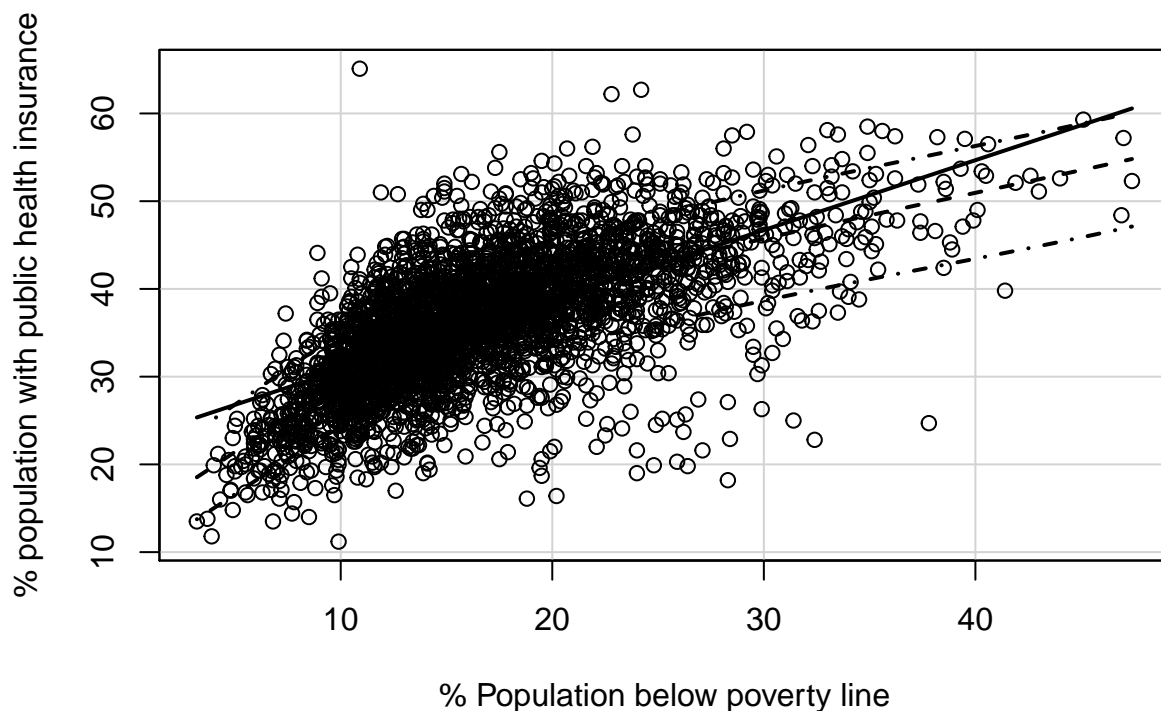


As expected, it presents an opposite behavior to median income. In this case, higher the poverty (and lower the income, as we previously saw), the lower private health insurance coverage.

Public Insurance Coverage

```
scatterplot(cancer.df$povertyPercent, cancer.df$PctPublicCoverage, ylab = "% population with public health insurance")
```

Public Coverage vs Poverty Percent

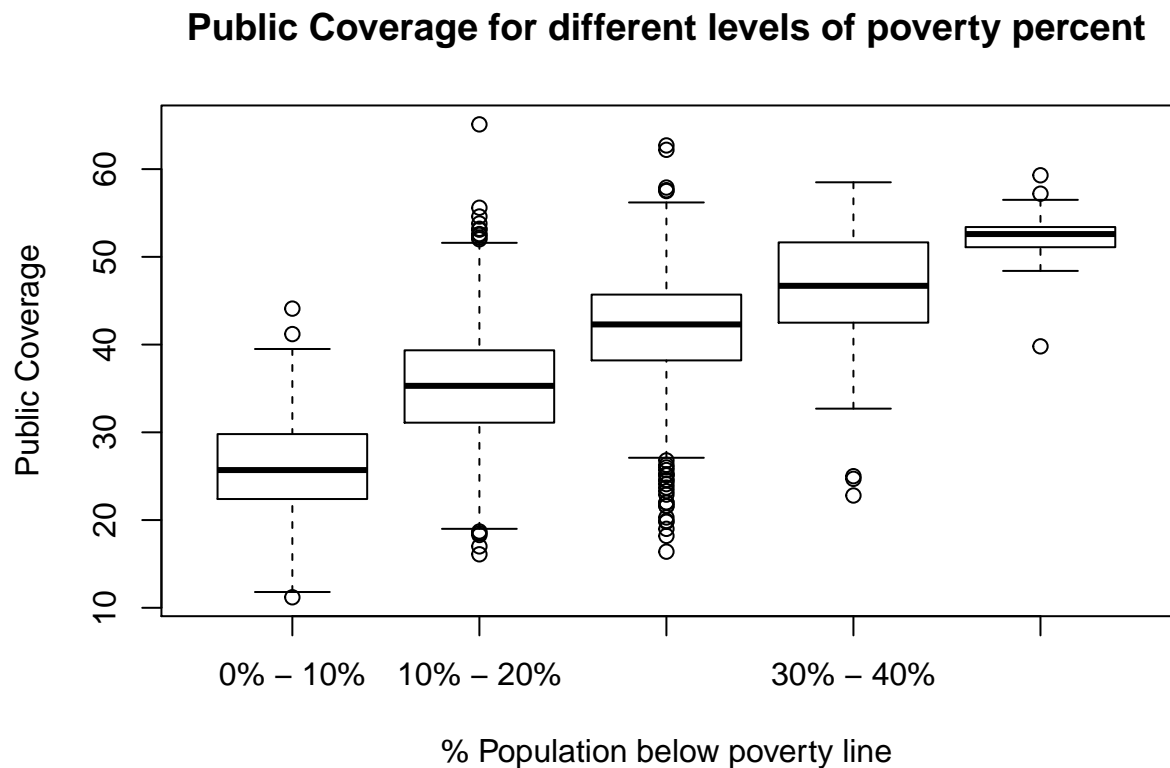


```
cor(cancer.df$povertyPercent, cancer.df$PctPublicCoverage)
```

```
## [1] 0.6511621
```

As expected by our previous analysis, the higher the poverty, more people rely on public health insurance. A deeper look into the levels of poverty vs public health coverage might provide us with better insights:

```
boxplot(cancer.df$PctPublicCoverage ~ cut(cancer.df$povertyPercent, right=FALSE, seq(0,50,10), labels = c(
  main = "Public Coverage for different levels of poverty percent",
  xlab = "% Population below poverty line", ylab = "Public Coverage")
```

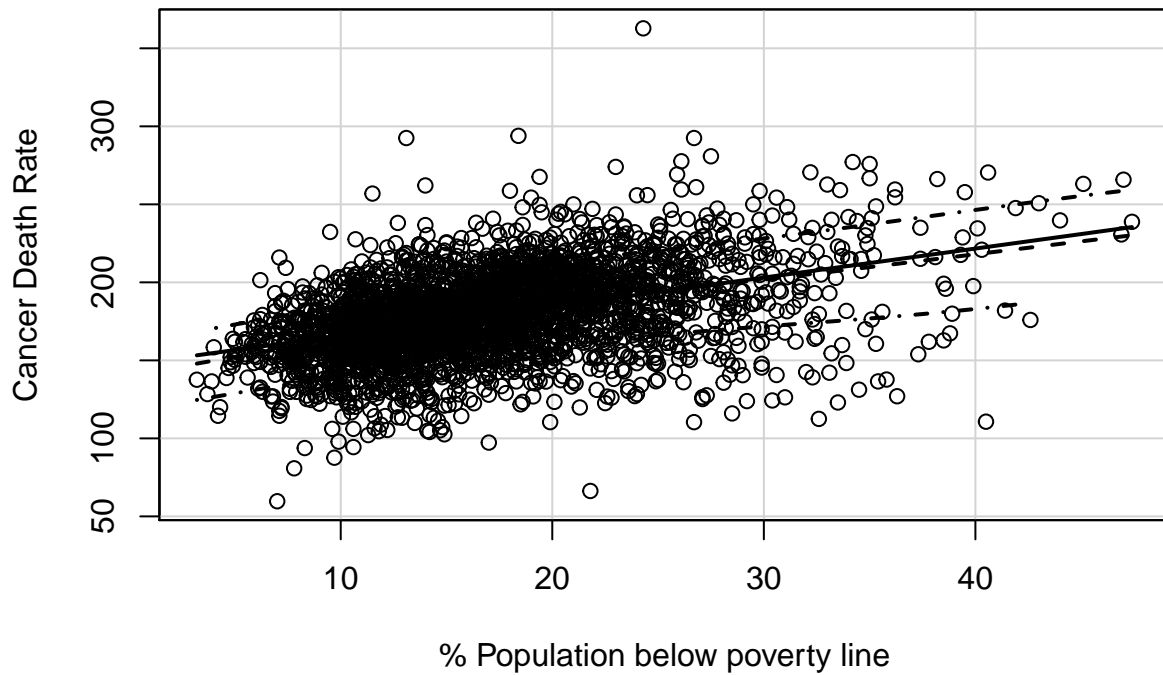


While analyzing the median income, we formulated the hypothesis that maybe the death rate is driven more due to socio-economic factors than to the percentage by type of health insurance coverage. We might confirm that by analyzing also the direct relation between death rate and poverty percent, with the result strenghtening or weakening this hypothesis.

deathRate

```
scatterplot(cancer.df$povertyPercent, cancer.df$deathRate, ylab = "Cancer Death Rate", xlab="% Population
```

Cancer Death Rate vs Poverty Percent



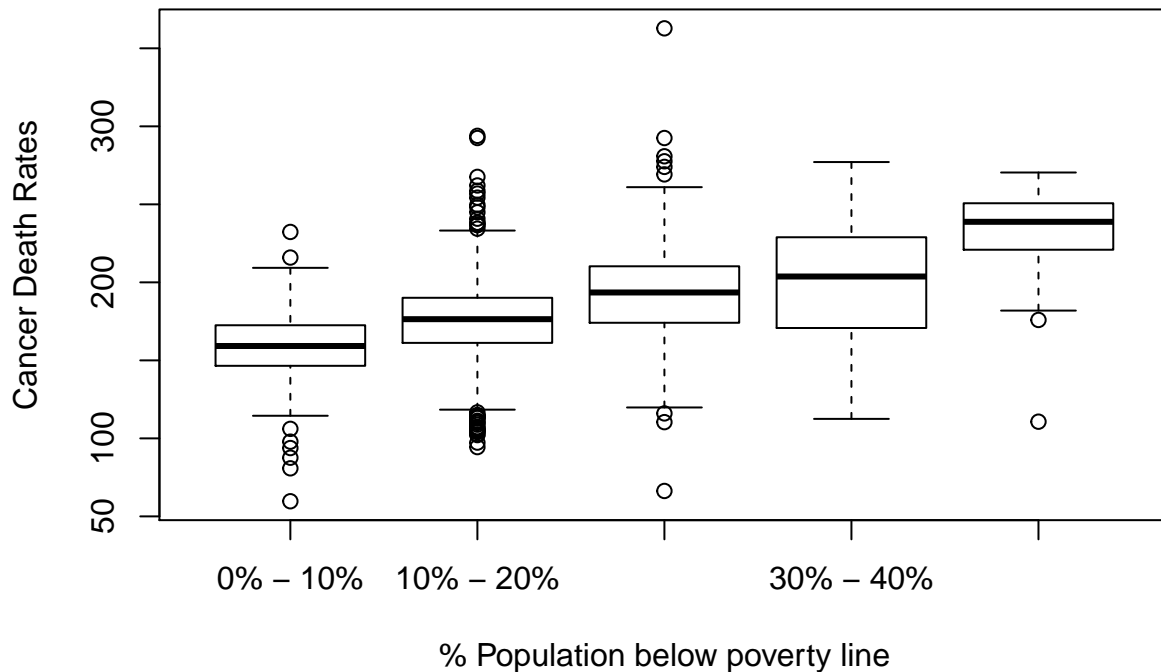
```
cor(cancer.df$povertyPercent, cancer.df$deathRate)
```

```
## [1] 0.429389
```

At first glance, we get a stronger positive relation of death rates and poverty percent than death rates and public health insurance coverage and death rates, indicating that we might be on the right track. To be more sure of it, we can make use of a boxplot by levels of poverty:

```
boxplot(cancer.df$deathRate ~ cut(cancer.df$povertyPercent, right=FALSE, seq(0,50,10), labels = c("0% - 10%", "10% - 20%", "20% - 30%", "30% - 40%", "40% - 50%")),  
  main = "Cancer Death Rates for different levels of poverty percent",  
  xlab = "% Population below poverty line", ylab = "Cancer Death Rates")
```

Cancer Death Rates for different levels of poverty percent



That all seems to confirm what we have seen analyzing the median income vs the death rates. Higher income / Lower poverty counties tend to have lower death rates.

Conclusion: insurance coverage per ce doesn't improve cancer mortality rates, however, better social economic conditions seems to do so

1. The opposite behavior of public health insurance coverage and private health insurance coverage to the death rates are most likely due to an underlying factor: social economic conditions.
2. Higher income populations tend to have lower cancer death rates, and with more money, more access to private health insurance.
3. Populations with higher percentage of poverty tend to have higher cancer death rates, and poverty conditions limitate the access to private health insurance coverage, being more dependent on the public alternative.
4. Therefore, there is stronger evidence that social economic factors (income, poverty) are stronger factors in explaining the cancer death rates than health insurance per ce, being that the coverage by type of health insurance is also probably affected by these factors, explaining their opposite behaviors with death rates.