

December 16, 2023

The results below are generated from an R script.

```
library(ggplot2)
communitySurvey <- read.csv("2014 American Community Survey.csv")
communitySurvey
```

##		Id	Id2	Geography	PopGroupID
## 1	05000000US01073	1073		Jefferson County, Alabama	1
## 2	05000000US04013	4013		Maricopa County, Arizona	1
## 3	05000000US04019	4019		Pima County, Arizona	1
## 4	05000000US06001	6001		Alameda County, California	1
## 5	05000000US06013	6013		Contra Costa County, California	1
## 6	05000000US06019	6019		Fresno County, California	1
## 7	05000000US06029	6029		Kern County, California	1
## 8	05000000US06037	6037		Los Angeles County, California	1
## 9	05000000US06059	6059		Orange County, California	1
## 10	05000000US06065	6065		Riverside County, California	1
## 11	05000000US06067	6067		Sacramento County, California	1
## 12	05000000US06071	6071		San Bernardino County, California	1
## 13	05000000US06073	6073		San Diego County, California	1
## 14	05000000US06075	6075		San Francisco County, California	1
## 15	05000000US06077	6077		San Joaquin County, California	1
## 16	05000000US06081	6081		San Mateo County, California	1
## 17	05000000US06085	6085		Santa Clara County, California	1
## 18	05000000US06097	6097		Sonoma County, California	1
## 19	05000000US06099	6099		Stanislaus County, California	1
## 20	05000000US06111	6111		Ventura County, California	1
## 21	05000000US08005	8005		Arapahoe County, Colorado	1
## 22	05000000US08031	8031		Denver County, Colorado	1
## 23	05000000US08041	8041		El Paso County, Colorado	1
## 24	05000000US08059	8059		Jefferson County, Colorado	1
## 25	05000000US09001	9001		Fairfield County, Connecticut	1
## 26	05000000US09003	9003		Hartford County, Connecticut	1
## 27	05000000US09009	9009		New Haven County, Connecticut	1
## 28	05000000US10003	10003		New Castle County, Delaware	1
## 29	05000000US11001	11001	District of Columbia	District of Columbia	1
## 30	05000000US12009	12009		Brevard County, Florida	1
## 31	05000000US12011	12011		Broward County, Florida	1
## 32	05000000US12031	12031		Duval County, Florida	1
## 33	05000000US12057	12057		Hillsborough County, Florida	1
## 34	05000000US12071	12071		Lee County, Florida	1
## 35	05000000US12086	12086		Miami-Dade County, Florida	1
## 36	05000000US12095	12095		Orange County, Florida	1
## 37	05000000US12099	12099		Palm Beach County, Florida	1
## 38	05000000US12103	12103		Pinellas County, Florida	1

## 39	0500000US12105	12105	Polk County, Florida	1
## 40	0500000US12127	12127	Volusia County, Florida	1
## 41	0500000US13067	13067	Cobb County, Georgia	1
## 42	0500000US13089	13089	DeKalb County, Georgia	1
## 43	0500000US13121	13121	Fulton County, Georgia	1
## 44	0500000US13135	13135	Gwinnett County, Georgia	1
## 45	0500000US15003	15003	Honolulu County, Hawaii	1
## 46	0500000US17031	17031	Cook County, Illinois	1
## 47	0500000US17043	17043	DuPage County, Illinois	1
## 48	0500000US17089	17089	Kane County, Illinois	1
## 49	0500000US17097	17097	Lake County, Illinois	1
## 50	0500000US17197	17197	Will County, Illinois	1
## 51	0500000US18097	18097	Marion County, Indiana	1
## 52	0500000US20091	20091	Johnson County, Kansas	1
## 53	0500000US20173	20173	Sedgwick County, Kansas	1
## 54	0500000US21111	21111	Jefferson County, Kentucky	1
## 55	0500000US24003	24003	Anne Arundel County, Maryland	1
## 56	0500000US24005	24005	Baltimore County, Maryland	1
## 57	0500000US24031	24031	Montgomery County, Maryland	1
## 58	0500000US24033	24033	Prince George's County, Maryland	1
## 59	0500000US24510	24510	Baltimore city, Maryland	1
## 60	0500000US25005	25005	Bristol County, Massachusetts	1
## 61	0500000US25009	25009	Essex County, Massachusetts	1
## 62	0500000US25017	25017	Middlesex County, Massachusetts	1
## 63	0500000US25021	25021	Norfolk County, Massachusetts	1
## 64	0500000US25023	25023	Plymouth County, Massachusetts	1
## 65	0500000US25025	25025	Suffolk County, Massachusetts	1
## 66	0500000US25027	25027	Worcester County, Massachusetts	1
## 67	0500000US26081	26081	Kent County, Michigan	1
## 68	0500000US26099	26099	Macomb County, Michigan	1
## 69	0500000US26125	26125	Oakland County, Michigan	1
## 70	0500000US26163	26163	Wayne County, Michigan	1
## 71	0500000US27053	27053	Hennepin County, Minnesota	1
## 72	0500000US27123	27123	Ramsey County, Minnesota	1
## 73	0500000US29095	29095	Jackson County, Missouri	1
## 74	0500000US29189	29189	St. Louis County, Missouri	1
## 75	0500000US31055	31055	Douglas County, Nebraska	1
## 76	0500000US32003	32003	Clark County, Nevada	1
## 77	0500000US34003	34003	Bergen County, New Jersey	1
## 78	0500000US34007	34007	Camden County, New Jersey	1
## 79	0500000US34013	34013	Essex County, New Jersey	1
## 80	0500000US34017	34017	Hudson County, New Jersey	1
## 81	0500000US34023	34023	Middlesex County, New Jersey	1
## 82	0500000US34025	34025	Monmouth County, New Jersey	1
## 83	0500000US34029	34029	Ocean County, New Jersey	1
## 84	0500000US34031	34031	Passaic County, New Jersey	1
## 85	0500000US34039	34039	Union County, New Jersey	1
## 86	0500000US35001	35001	Bernalillo County, New Mexico	1
## 87	0500000US36005	36005	Bronx County, New York	1
## 88	0500000US36029	36029	Erie County, New York	1
## 89	0500000US36047	36047	Kings County, New York	1
## 90	0500000US36055	36055	Monroe County, New York	1
## 91	0500000US36059	36059	Nassau County, New York	1
## 92	0500000US36061	36061	New York County, New York	1

## 93	0500000US36081	36081	Queens County, New York	1
## 94	0500000US36103	36103	Suffolk County, New York	1
## 95	0500000US36119	36119	Westchester County, New York	1
## 96	0500000US37081	37081	Guilford County, North Carolina	1
## 97	0500000US37119	37119	Mecklenburg County, North Carolina	1
## 98	0500000US37183	37183	Wake County, North Carolina	1
## 99	0500000US39035	39035	Cuyahoga County, Ohio	1
## 100	0500000US39049	39049	Franklin County, Ohio	1
## 101	0500000US39061	39061	Hamilton County, Ohio	1
## 102	0500000US39113	39113	Montgomery County, Ohio	1
## 103	0500000US39153	39153	Summit County, Ohio	1
## 104	0500000US40109	40109	Oklahoma County, Oklahoma	1
## 105	0500000US40143	40143	Tulsa County, Oklahoma	1
## 106	0500000US41051	41051	Multnomah County, Oregon	1
## 107	0500000US41067	41067	Washington County, Oregon	1
## 108	0500000US42003	42003	Allegheny County, Pennsylvania	1
## 109	0500000US42017	42017	Bucks County, Pennsylvania	1
## 110	0500000US42029	42029	Chester County, Pennsylvania	1
## 111	0500000US42045	42045	Delaware County, Pennsylvania	1
## 112	0500000US42071	42071	Lancaster County, Pennsylvania	1
## 113	0500000US42091	42091	Montgomery County, Pennsylvania	1
## 114	0500000US42101	42101	Philadelphia County, Pennsylvania	1
## 115	0500000US44007	44007	Providence County, Rhode Island	1
## 116	0500000US47037	47037	Davidson County, Tennessee	1
## 117	0500000US47157	47157	Shelby County, Tennessee	1
## 118	0500000US48029	48029	Bexar County, Texas	1
## 119	0500000US48085	48085	Collin County, Texas	1
## 120	0500000US48113	48113	Dallas County, Texas	1
## 121	0500000US48121	48121	Denton County, Texas	1
## 122	0500000US48141	48141	El Paso County, Texas	1
## 123	0500000US48157	48157	Fort Bend County, Texas	1
## 124	0500000US48201	48201	Harris County, Texas	1
## 125	0500000US48215	48215	Hidalgo County, Texas	1
##	POPGROUP.display.label	RacesReported	HSDegree	BachDegree
## 1	Total population	660793	89.1	30.5
## 2	Total population	4087191	86.8	30.2
## 3	Total population	1004516	88.0	30.8
## 4	Total population	1610921	86.9	42.8
## 5	Total population	1111339	88.8	39.7
## 6	Total population	965974	73.6	19.7
## 7	Total population	874589	74.5	15.4
## 8	Total population	10116705	77.5	30.3
## 9	Total population	3145515	84.6	38.0
## 10	Total population	2329271	80.6	20.7
## 11	Total population	1482026	86.8	28.9
## 12	Total population	2112619	78.6	18.9
## 13	Total population	3263431	86.6	37.1
## 14	Total population	852469	88.1	54.2
## 15	Total population	715597	77.6	18.3
## 16	Total population	758581	88.1	47.5
## 17	Total population	1894605	87.4	48.4
## 18	Total population	500292	87.6	34.8
## 19	Total population	531997	78.4	17.0
## 20	Total population	846178	83.6	31.6

## 21	Total population	618821	91.9	40.9
## 22	Total population	663862	85.5	44.3
## 23	Total population	663519	92.8	36.5
## 24	Total population	558503	94.1	42.0
## 25	Total population	945438	89.8	46.7
## 26	Total population	897985	89.3	36.8
## 27	Total population	861277	89.5	34.5
## 28	Total population	552778	90.1	35.8
## 29	Total population	658893	90.2	55.0
## 30	Total population	556885	91.6	27.2
## 31	Total population	1869235	88.4	30.5
## 32	Total population	897698	89.0	26.1
## 33	Total population	1316298	87.3	29.8
## 34	Total population	679513	86.3	26.5
## 35	Total population	2662874	80.9	26.6
## 36	Total population	1253001	87.9	31.4
## 37	Total population	1397710	87.7	33.0
## 38	Total population	938098	90.1	29.5
## 39	Total population	634638	84.9	19.7
## 40	Total population	507531	88.9	22.5
## 41	Total population	730981	90.3	43.7
## 42	Total population	722161	88.4	41.7
## 43	Total population	996319	91.3	49.2
## 44	Total population	877922	88.0	35.4
## 45	Total population	991788	91.8	32.6
## 46	Total population	5246456	85.5	36.2
## 47	Total population	932708	92.3	48.0
## 48	Total population	527306	82.9	32.6
## 49	Total population	705186	90.3	44.0
## 50	Total population	685419	90.7	33.1
## 51	Total population	934243	85.0	28.8
## 52	Total population	574272	95.5	52.8
## 53	Total population	508803	88.8	30.7
## 54	Total population	760026	88.5	31.6
## 55	Total population	560133	91.9	38.8
## 56	Total population	826925	90.4	37.2
## 57	Total population	1030447	90.9	58.5
## 58	Total population	904430	85.5	31.0
## 59	Total population	622793	84.4	30.0
## 60	Total population	554194	82.5	25.7
## 61	Total population	769091	89.1	38.9
## 62	Total population	1570315	92.3	52.3
## 63	Total population	692254	94.1	51.9
## 64	Total population	507022	92.2	34.1
## 65	Total population	767254	83.9	42.3
## 66	Total population	813475	90.1	34.6
## 67	Total population	629237	89.1	33.7
## 68	Total population	860112	89.3	23.9
## 69	Total population	1237868	93.6	44.8
## 70	Total population	1764804	84.9	22.1
## 71	Total population	1212064	93.2	47.3
## 72	Total population	532655	89.9	40.9
## 73	Total population	683191	90.0	29.5
## 74	Total population	1001876	93.2	42.8

```
## 75      Total population      543244      88.2      36.3
## 76      Total population     2069681      84.5      22.7
## 77      Total population      933572      91.5      46.2
## 78      Total population      511038      88.3      31.3
## 79      Total population      795723      85.5      32.7
## 80      Total population      669115      83.4      38.2
## 81      Total population      836297      89.1      41.0
## 82      Total population      629279      93.1      43.7
## 83      Total population      586301      91.7      28.6
## 84      Total population      508856      83.8      28.6
## 85      Total population      552939      86.2      33.0
## 86      Total population      675551      88.0      32.7
## 87      Total population     1438159      70.5      19.3
## 88      Total population      922835      90.6      31.3
## 89      Total population     2621793      80.0      34.3
## 90      Total population      749857      90.3      35.9
## 91      Total population     1358627      90.7      43.2
## 92      Total population     1636268      86.8      59.9
## 93      Total population     2321580      80.4      29.8
## 94      Total population     1502968      89.8      34.0
## 95      Total population      972634      87.4      47.1
## 96      Total population      512119      89.0      33.3
## 97      Total population     1012539      89.5      43.0
## 98      Total population      998691      92.4      49.2
## 99      Total population     1259828      88.1      31.0
## 100     Total population     1231393      90.0      38.0
## 101     Total population      806631      90.5      35.6
## 102     Total population      533116      89.7      25.7
## 103     Total population      541943      91.1      30.3
## 104     Total population      766215      86.8      30.6
## 105     Total population      629598      88.6      30.7
## 106     Total population      776712      91.1      41.6
## 107     Total population      562998      90.2      39.7
## 108     Total population     1231255      93.9      37.7
## 109     Total population      626685      93.9      37.7
## 110     Total population      512784      92.3      49.3
## 111     Total population      562960      91.5      36.3
## 112     Total population      533320      84.9      26.0
## 113     Total population      816857      93.7      47.3
## 114     Total population     1560297      82.6      26.0
## 115     Total population      631974      82.0      25.2
## 116     Total population      668347      86.7      37.3
## 117     Total population      938803      87.4      29.9
## 118     Total population     1855866      83.0      26.3
## 119     Total population      885241      93.7      50.0
## 120     Total population     2518638      77.6      29.1
## 121     Total population      753363      91.9      41.5
## 122     Total population      833487      75.8      21.1
## 123     Total population      685345      88.6      44.1
## 124     Total population     4441370      79.8      29.7
## 125     Total population      831073      62.2      17.9
## [ reached 'max' / getOption("max.print") -- omitted 11 rows ]

lapply(communitySurvey, class)
```

```
## $Id
## [1] "character"
##
## $Id2
## [1] "integer"
##
## $Geography
## [1] "character"
##
## $PopGroupID
## [1] "integer"
##
## $POPGROUP.display.label
## [1] "character"
##
## $RacesReported
## [1] "integer"
##
## $HSDegree
## [1] "numeric"
##
## $BachDegree
## [1] "numeric"

# ID is a character. It is a unique alpha/numeric character used to identify the city and state being s
# ID2 is an integer. It is a shortened version of ID which only included the last 5 digits of the ID ch
# Geography is a character. It shows the city and state in written form.
# PopGroupID is an integer. It is an identified for the population group.
# POP.GROUP.display.label is a character. This labels what population we are talking about. All of thes
# RacesReported is an integer. This shows the number of races reported in the city/state.
# HSDegree is numeric. It shows the percentage of the population that acquired a high school degree.
# BachDegree is numeric. It shows the percentage of the population that acquired a bachelor's degree.

# Run the following functions and provide the results: str(); nrow(); ncol()
str(communitySurvey)

## 'data.frame': 136 obs. of 8 variables:
## $ Id : chr "0500000US01073" "0500000US04013" "0500000US04019" "0500000US06001" ...
## $ Id2 : int 1073 4013 4019 6001 6013 6019 6029 6037 6059 6065 ...
## $ Geography : chr "Jefferson County, Alabama" "Maricopa County, Arizona" "Pima County,
## $ PopGroupID : int 1 1 1 1 1 1 1 1 1 1 ...
## $ POPGROUP.display.label: chr "Total population" "Total population" "Total population" "Total popul
## $ RacesReported : int 660793 4087191 1004516 1610921 1111339 965974 874589 10116705 3145515
## $ HSDegree : num 89.1 86.8 88 86.9 88.8 73.6 74.5 77.5 84.6 80.6 ...
## $ BachDegree : num 30.5 30.2 30.8 42.8 39.7 19.7 15.4 30.3 38 20.7 ...

nrow(communitySurvey)

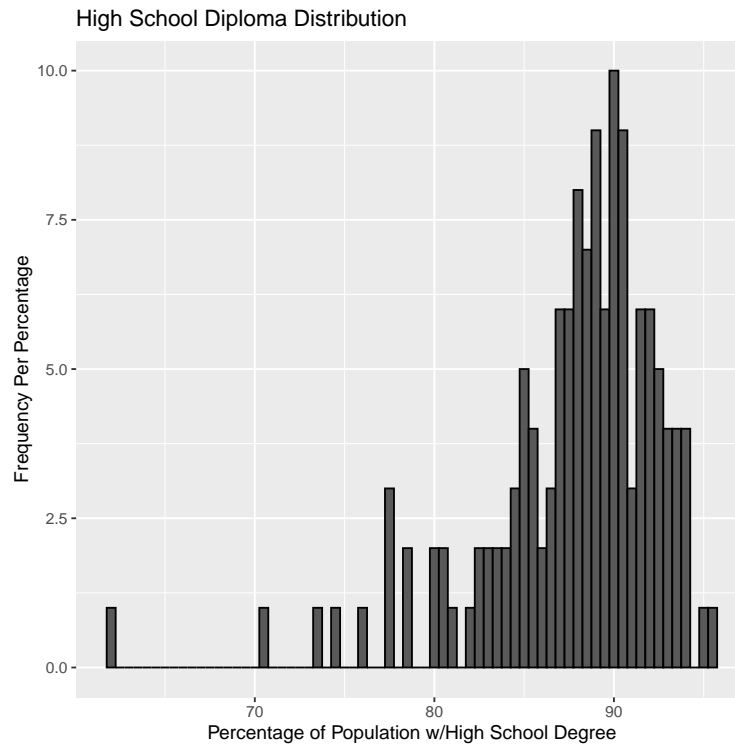
## [1] 136

ncol(communitySurvey)

## [1] 8

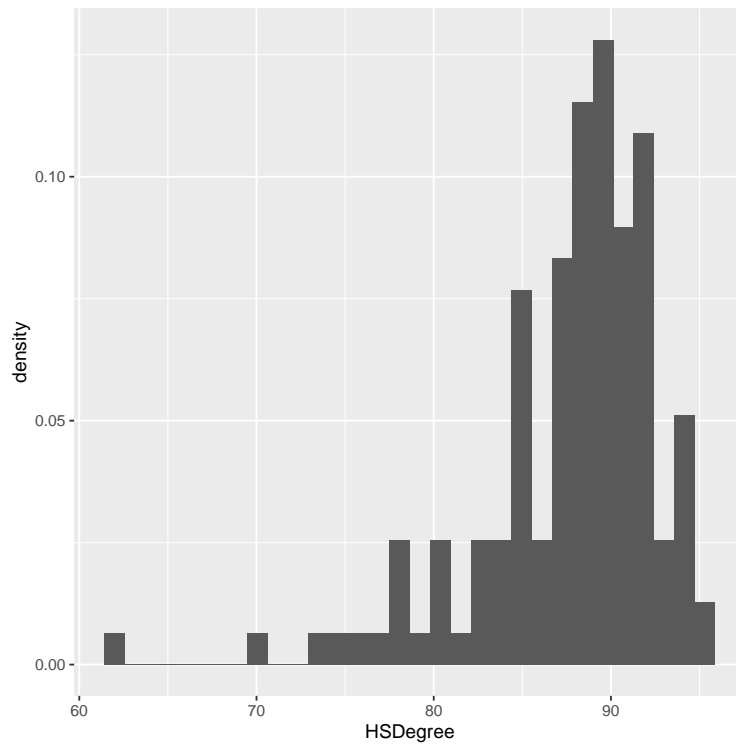
# Create a Histogram of the HSDegree variable using the ggplot2 package.
```

```
# Set a bin size for the Histogram that you think best visualizes the data (the bin size will determine how many bars are shown)
# Include a Title and appropriate X/Y axis labels on your Histogram Plot.
communitySurveyHistogram <- ggplot(communitySurvey, aes(HSDegree))
communitySurveyHistogram + geom_histogram(color = "Black", binwidth = 0.5) + labs(title = "High School Diploma Distribution")
```



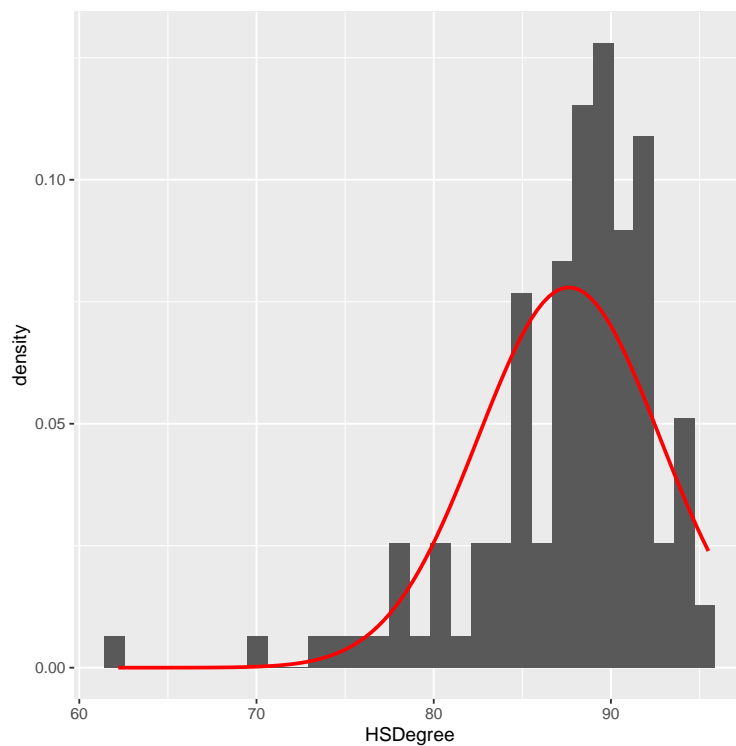
```
# Answer the following questions based on the Histogram produced.
# 1) Based on what you see in this histogram, is the data distribution unimodal?
# Based on the histogram created, the data distribution is unimodal as there is a distinct peak at about 90.5.
# 2) Is it approximately symmetrical?
# It is not approximately symmetrical since even though it has a distinct peak, the two tails on either side are not equal.
# 3) Is it approximately bell-shaped?
# The histogram is not bell-shaped.
# 4) Is it approximately normal?
# The histogram is not normally distributed.
# 5) If not normal, is the distribution skewed? If so, in which direction?
# The distribution is skewed. It presents itself with a left-skew.
# 6) Include a normal curve to the Histogram that you plotted.
ggplot(communitySurvey, aes(HSDegree)) + geom_histogram(aes(y = ..density..))

## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```



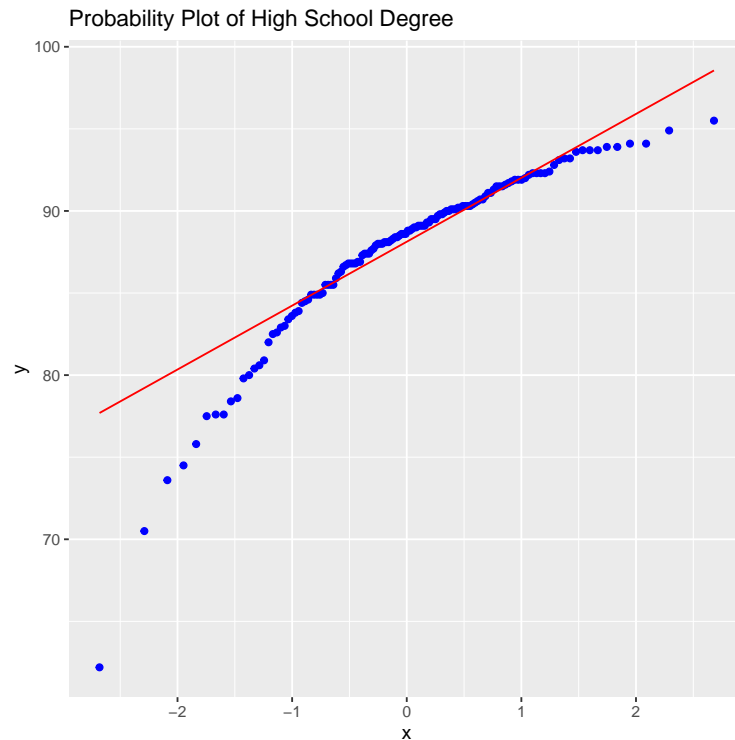
```
ggplot(communitySurvey, aes(HSDegree)) + geom_histogram(aes(y = ..density..)) +
  stat_function(fun = dnorm, args = list(mean = mean(communitySurvey$HSDegree),
    sd = sd(communitySurvey$HSDegree)), color = "Red", size = 1)

## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```




```
# 7) Explain whether a normal distribution can accurately be used as a model for this data.
# curve(dnorm(communitySurvey$HSDegree, mean(communitySurvey$HSDegree), sd(communitySurvey$HSDegree), a
# Since the histogram has a negative skew, a normal distribution cannot accurately be used as a model for

# Create a Probability Plot of the HSDegree variable.
ggplot(communitySurvey, aes(sample = HSDegree)) + geom_qq(color = "Blue") + geom_qq_line(color = "Red") +
```



```
# Answer the following questions based on the Probability Plot:
# 1) Based on what you see in this probability plot, is the distribution approximately normal? Explain
# Based on the probability plot, the distribution is not approximately normal since it does not follow
# 2) If not normal, is the distribution skewed? If so, in which direction? Explain how you know.
# The distribution is skewed to the left since the plotted points bend down and to the right of the norm

# Now that you have looked at this data visually for normality, you will now quantify normality with nu
library(pastecs)
stat.desc(communitySurvey$HSDegree)

##      nbr.val    nbr.null    nbr.na      min      max      range
## 1.360000e+02 0.000000e+00 0.000000e+00 6.220000e+01 9.550000e+01 3.330000e+01
##      sum      median      mean    SE.mean CI.mean.0.95      var
## 1.191800e+04 8.870000e+01 8.763235e+01 4.388598e-01 8.679296e-01 2.619332e+01
##      std.dev      coef.var
## 5.117941e+00 5.840241e-02

library(moments)
skewness(communitySurvey$HSDegree)

## [1] -1.69341

kurtosis(communitySurvey$HSDegree)
```

```
## [1] 7.462191

mean <- mean(communitySurvey$HSDegree)
sd <- sd(communitySurvey$HSDegree)
z_scores <- (communitySurvey$HSDegree - mean / sd)
z_scores

## [1] 71.97742 69.67742 70.87742 69.77742 71.67742 56.47742 57.37742 60.37742 67.47742
## [10] 63.47742 69.67742 61.47742 69.47742 70.97742 60.47742 70.97742 70.27742 70.47742
## [19] 61.27742 66.47742 74.77742 68.37742 75.67742 76.97742 72.67742 72.17742 72.37742
## [28] 72.97742 73.07742 74.47742 71.27742 71.87742 70.17742 69.17742 63.77742 70.77742
## [37] 70.57742 72.97742 67.77742 71.77742 73.17742 71.27742 74.17742 70.87742 74.67742
## [46] 68.37742 75.17742 65.77742 73.17742 73.57742 67.87742 78.37742 71.67742 71.37742
## [55] 74.77742 73.27742 73.77742 68.37742 67.27742 65.37742 71.97742 75.17742 76.97742
## [64] 75.07742 66.77742 72.97742 71.97742 72.17742 76.47742 67.77742 76.07742 72.77742
## [73] 72.87742 76.07742 71.07742 67.37742 74.37742 71.17742 68.37742 66.27742 71.97742
## [82] 75.97742 74.57742 66.67742 69.07742 70.87742 53.37742 73.47742 62.87742 73.17742
## [91] 73.57742 69.67742 63.27742 72.67742 70.27742 71.87742 72.37742 75.27742 70.97742
## [100] 72.87742 73.37742 72.57742 73.97742 69.67742 71.47742 73.97742 73.07742 76.77742
## [109] 76.77742 75.17742 74.37742 67.77742 76.57742 65.47742 64.87742 69.57742 70.27742
## [118] 65.87742 76.57742 60.47742 74.77742 58.67742 71.47742 62.67742 45.07742 68.77742
## [127] 67.77742 71.47742 72.37742 76.57742 74.37742 75.17742 73.17742 74.87742 77.77742
## [136] 69.77742

# In several sentences provide an explanation of the result produced for skew, kurtosis, and z-scores. I
# Skewness is a statistical measure that is used to show whether a distribution is distorted or asymmetr
# Kurtosis measures the degree of peakedness in a distribution. A kurtosis value of 7.462191 indicates
# Z-scores are a standardized measure of how far a data point is from the mean, expressed in terms of th
# I attempted to get z-scores but don't believe I returned the correct values as the numbers are too lar
```

The R session information (including the OS info, R version and all packages used):

```
sessionInfo()

## R version 4.3.2 (2023-10-31)
## Platform: x86_64-apple-darwin20 (64-bit)
## Running under: macOS Ventura 13.5.1
##
## Matrix products: default
## BLAS: /System/Library/Frameworks/Accelerate.framework/Versions/A/Frameworks/vecLib.framework/Versions/1.17.0
## LAPACK: /Library/Frameworks/R.framework/Versions/4.3-x86_64/Resources/lib/libRlapack.dylib; LAPACK v
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
##
## time zone: America/New_York
## tzcode source: internal
##
## attached base packages:
## [1] splines stats graphics grDevices utils datasets methods base
##
## other attached packages:
## [1] moments_0.14.1 pastecs_1.3.21 effects_4.2-2 RcmdrMisc_2.9-1 sandwich_3.1-0
## [6] car_3.1-2 carData_3.0-5 ggthemes_5.0.0 scales_1.2.1 lubridate_1.9.3
## [11] shiny_1.8.0 DataEditR_0.1.5 ggplot2_3.4.4 jsonlite_1.8.8
```

```
##
## loaded via a namespace (and not attached):
## [1] DBI_1.1.3          gridExtra_2.3      tcltk_4.3.2        readxl_1.4.3
## [5] rlang_1.1.2        magrittr_2.0.3     e1071_1.7-14       compiler_4.3.2
## [9] mgcv_1.9-0         vctrs_0.6.4        stringr_1.5.1      crayon_1.5.2
## [13] pkgconfig_2.0.3    fastmap_1.1.1      backports_1.4.1    ellipsis_0.3.2
## [17] fontawesome_0.5.2  labeling_0.4.3     utf8_1.2.4         promises_1.2.1
## [21] rmarkdown_2.25     haven_2.5.4        nloptr_2.0.3       purrr_1.0.2
## [25] xfun_0.41          cachem_1.0.8       highr_0.10         later_1.3.1
## [29] cluster_2.1.4      R6_2.5.1           bslib_0.6.0        stringi_1.8.2
## [33] boot_1.3-28.1      rpart_4.1.21       jquerylib_0.1.4    cellranger_1.1.0
## [37] Rcpp_1.0.11        knitr_1.45         zoo_1.8-12         base64enc_0.1-3
## [41] httpuv_1.6.13      Matrix_1.6-1.1     nnet_7.3-19        timechange_0.2.0
## [45] tidyselect_1.2.0   rstudioapi_0.15.0  abind_1.4-5        yaml_2.3.7
## [49] miniUI_0.1.1.1     lattice_0.22-5     tibble_3.2.1       withr_2.5.2
## [53] evaluate_0.23      foreign_0.8-85     survival_3.5-7     proxy_0.4-27
## [57] survey_4.2-1       pillar_1.9.0       checkmate_2.3.1    nortest_1.0-4
## [61] insight_0.19.7     shinyjs_2.1.0      generics_0.1.3     hms_1.1.3
## [65] munsell_0.5.0      minqa_1.2.6        xtable_1.8-4       class_7.3-22
## [69] glue_1.6.2         Hmisc_5.1-1        tools_4.3.2        data.table_1.14.8
## [73] lme4_1.1-35.1      forcats_1.0.0      fs_1.6.3           grid_4.3.2
## [77] mitools_2.4        shinyBS_0.61.1     colorspace_2.1-0   nlme_3.1-163
## [81] htmlTable_2.4.2    Formula_1.2-5      cli_3.6.1          rhandsontable_0.3.8
## [85] fansi_1.0.5        tcltk2_1.2-11     viridisLite_0.4.2  dplyr_1.1.4
## [89] gtable_0.3.4       relimp_1.0-5       sass_0.4.7         digest_0.6.33
## [93] htmlwidgets_1.6.3  farver_2.1.1       memoise_2.0.1      htmltools_0.5.7
## [97] lifecycle_1.0.4    mime_0.12          MASS_7.3-60

Sys.time()

## [1] "2023-12-16 21:32:10 EST"
```