

initial_analysis

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```
library(dplyr)
```

```
##  
## Attaching package: 'dplyr'  
  
## The following objects are masked from 'package:stats':  
##  
##   filter, lag  
  
## The following objects are masked from 'package:base':  
##  
##   intersect, setdiff, setequal, union
```

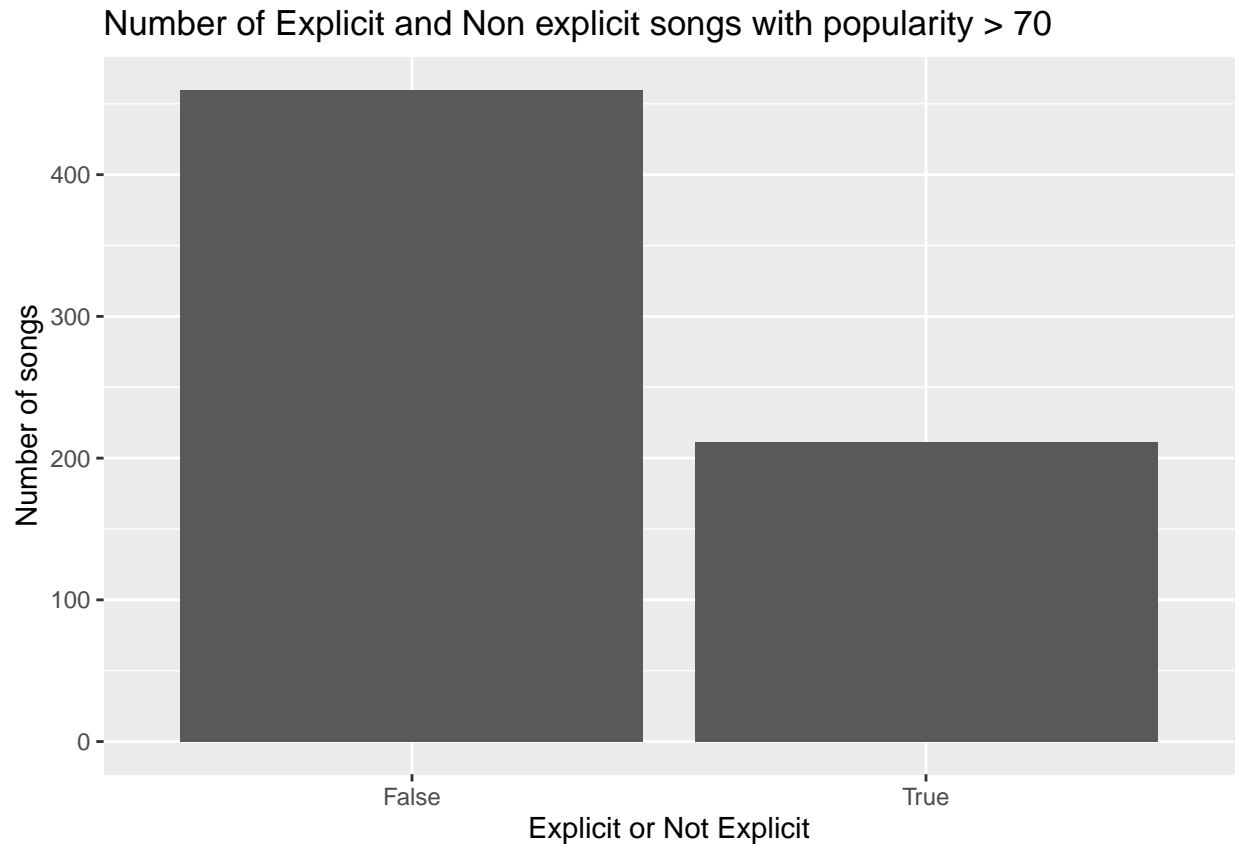
```
library(ggplot2)  
library(stringr)  
songs <- read.csv("data/songs_normalize.csv")
```

```
over70 <- filter(songs, popularity > 70)  
#over70  
  
Drake <- filter(songs, artist == "Drake")  
head(arrange(Drake, desc(popularity)))
```

```
##   artist                                song duration_ms explicit year  
## 1 Drake                                One Dance      173986     False 2016  
## 2 Drake                                God's Plan     198973      True 2018  
## 3 Drake                                Hotline Bling    267066     False 2016  
## 4 Drake                                Nonstop         238614      True 2018  
## 5 Drake                                Nice For What   210746      True 2018  
## 6 Drake Money In The Grave (Drake ft. Rick Ross) 205426      True 2019  
## popularity danceability energy key loudness mode speechiness acoustictness  
## 1          84          0.792 0.625 1   -5.609 1      0.0536      0.00776  
## 2          81          0.754 0.449 7   -9.211 1      0.1090      0.03320  
## 3          77          0.891 0.628 2   -7.863 1      0.0551      0.00258  
## 4          77          0.912 0.412 7   -8.074 1      0.1230      0.01650  
## 5          77          0.585 0.909 8   -6.474 1      0.0707      0.08910  
## 6          76          0.831 0.502 10  -4.045 0      0.0460      0.10100  
## instrumentalness liveness valence tempo genre  
## 1          1.80e-03 0.3290 0.370 103.967 hip hop, pop, R&B
```

```
## 2      8.29e-05  0.5520  0.357  77.169 hip hop, pop, R&B
## 3      1.90e-04  0.0504  0.552 134.966 hip hop, pop, R&B
## 4      1.26e-02  0.1040  0.423 154.983 hip hop, pop, R&B
## 5      9.70e-05  0.1190  0.758  93.372 hip hop, pop, R&B
## 6      0.00e+00  0.1220  0.101 100.541 hip hop, pop, R&B
```

```
ggplot(data = over70, aes(x = explicit)) + geom_bar() + labs(x = "Explicit or Not Explicit", y = "Number of songs")
```



```
notExplicit <- filter(songs, explicit== "False")
#notExplicit

top20 <- slice_max(songs, order_by = popularity, n = 20)
#top20

selected <- select(songs, popularity, song, danceability, energy, explicit, acousticness, liveness)
#selected

biggerSongs <- mutate(songs,
  duration_s = duration_ms/1000,
  duration_minutes = duration_s/60)
head(biggerSongs)
```

```
##          artist          song duration_ms explicit year popularity
## 1 Britney Spears Oops!...I Did It Again    211160    False 2000         77
## 2 blink-182    All The Small Things    167066    False 1999         79
```

```
## 3      Faith Hill          Breathe      250546      False 1999      66
## 4      Bon Jovi           It's My Life      224493      False 2000      78
## 5      *NSYNC             Bye Bye Bye      200560      False 2000      65
## 6      Sisqo              Thong Song       253733      True 1999      69
##      danceability energy key loudness mode speechiness acoustictness
## 1      0.751 0.834 1 -5.444 0 0.0437 0.3000
## 2      0.434 0.897 0 -4.918 1 0.0488 0.0103
## 3      0.529 0.496 7 -9.007 1 0.0290 0.1730
## 4      0.551 0.913 0 -4.063 0 0.0466 0.0263
## 5      0.614 0.928 8 -4.806 0 0.0516 0.0408
## 6      0.706 0.888 2 -6.959 1 0.0654 0.1190
##      instrumentalness liveness valence tempo genre duration_s
## 1      1.77e-05 0.3550 0.894 95.053 pop 211.160
## 2      0.00e+00 0.6120 0.684 148.726 rock, pop 167.066
## 3      0.00e+00 0.2510 0.278 136.859 pop, country 250.546
## 4      1.35e-05 0.3470 0.544 119.992 rock, metal 224.493
## 5      1.04e-03 0.0845 0.879 172.656 pop 200.560
## 6      9.64e-05 0.0700 0.714 121.549 hip hop, pop, R&B 253.733
##      duration_minutes
## 1      3.519333
## 2      2.784433
## 3      4.175767
## 4      3.741550
## 5      3.342667
## 6      4.228883
```

```
shortestSongs <- arrange(biggerSongs, duration_minutes)
longestSongs <- arrange(biggerSongs, desc(duration_minutes))
shortestModified <- select(longestSongs, popularity, song, duration_minutes)
#shortestModified
longestModified <- select(longestSongs, popularity, song, duration_minutes)
#longestModified

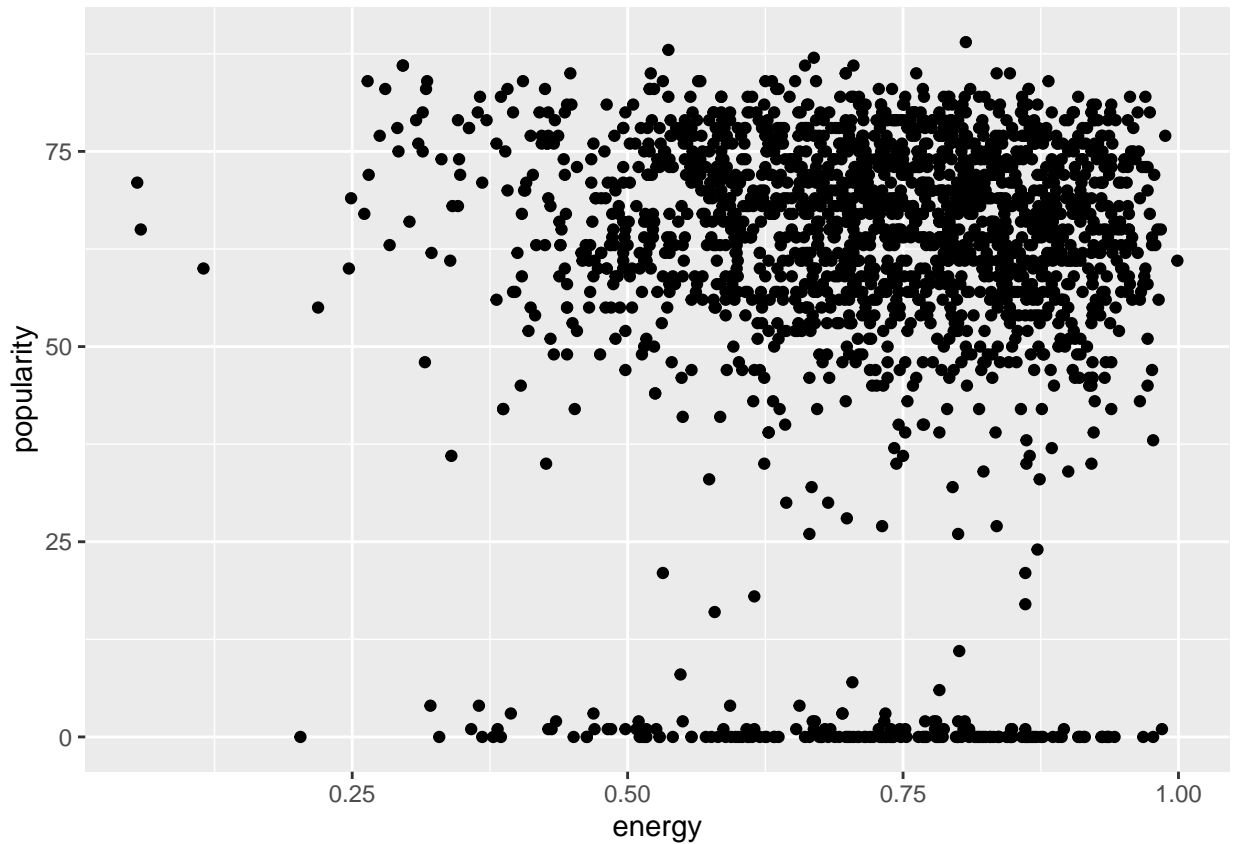
summarized <- summarise(songs, mean_popularity = mean(popularity, na.rm = T))
#summarized
numOfPopularity <- summarise(group_by(songs, popularity), count = n()) # n()
#numOfPopularity
```

```
## Practice subsetting data
# use a combination of filter, select, mutate, arrange, summarise, group_by, sample, and/or slice
# create a visulaization using your new subset of data
mostEnergetic <- filter(songs, energy > 0.50)
#mostEnergetic

arrangedEnergetic <- arrange(mostEnergetic, desc(popularity))
head(select(arrangedEnergetic, artist, song, popularity, energy))
```

```
##      artist      song popularity energy
## 1 The Neighbourhood  Sweater Weather      89 0.807
## 2      Tom Odell      Another Love      88 0.537
## 3      Eminem      Without Me      87 0.669
## 4      Eminem The Real Slim Shady      86 0.661
## 5      WILLOW      Wait a Minute!      86 0.705
## 6      Eminem      'Till I Collapse      85 0.847
```

```
ggplot(data = songs, aes(x = energy, y = popularity)) + geom_point()
```



```
head(songs %>%
  group_by(popularity) %>%
  sample_n(1))
```

```
## # A tibble: 6 x 18
## # Groups:   popularity [6]
##   artist  song  duration_ms explicit  year popularity danceability energy  key
##   <chr>   <chr>      <int> <chr>    <int>      <int>      <dbl>  <dbl> <int>
## 1 Astrid S Hurt~    208728 False    2016         0      0.672  0.589     7
## 2 Lil Way~ 6 Fo~    248586 True     2011         1      0.364  0.752     2
## 3 Adele    Set ~    242973 False    2011         2      0.603  0.67      2
## 4 The Whi~ Seve~    231920 False    2003         3      0.741  0.469     4
## 5 Baby Ba~ Baby~    219920 True     2005         4      0.899  0.365     9
## 6 Avicii   Wake~    247426 False    2013         6      0.532  0.783     2
## # ... with 9 more variables: loudness <dbl>, mode <int>, speechiness <dbl>,
## #   acoustictness <dbl>, instrumentaltness <dbl>, liveness <dbl>, valence <dbl>,
## #   tempo <dbl>, genre <chr>
```

```
head(songs %>%
  group_by(year) %>%
  sample_n(1))
```

```
## # A tibble: 6 x 18
## # Groups:   year [6]
##   artist    song duration_ms explicit  year popularity danceability energy  key
##   <chr>    <chr>      <int> <chr>   <int>      <int>      <dbl>  <dbl> <int>
## 1 Missy E~ Hot ~      215466 True    1998         49      0.727  0.445    1
## 2 Dido     Than~      218360 False   1999         73      0.725  0.583    1
## 3 Eminem   Stan      404106 True    2000         83      0.78   0.768    6
## 4 S Club 7 Don'~      233626 False   2001         63      0.822  0.672    7
## 5 JAY-Z    Excu~      281240 True    2002         56      0.714  0.862    6
## 6 Three D~ I Ha~      231480 False   2003         72      0.498  0.83     6
## # ... with 9 more variables: loudness <dbl>, mode <int>, speechiness <dbl>,
## #   acoustictness <dbl>, instrumentatness <dbl>, liveness <dbl>, valence <dbl>,
## #   tempo <dbl>, genre <chr>
```

```
mu <- mean(songs$loudness)
sig <- sd(songs$loudness)
iqr <- IQR(songs$loudness)
q1 <- as.numeric(quantile(songs$loudness, 0.25))
q3 <- as.numeric(quantile(songs$loudness, 0.75))
mu - 3*sig #min1
```

```
## [1] -11.31288
```

```
mu + 3*sig #max1
```

```
## [1] 0.2880115
```

```
q1 - iqr *1.5
```

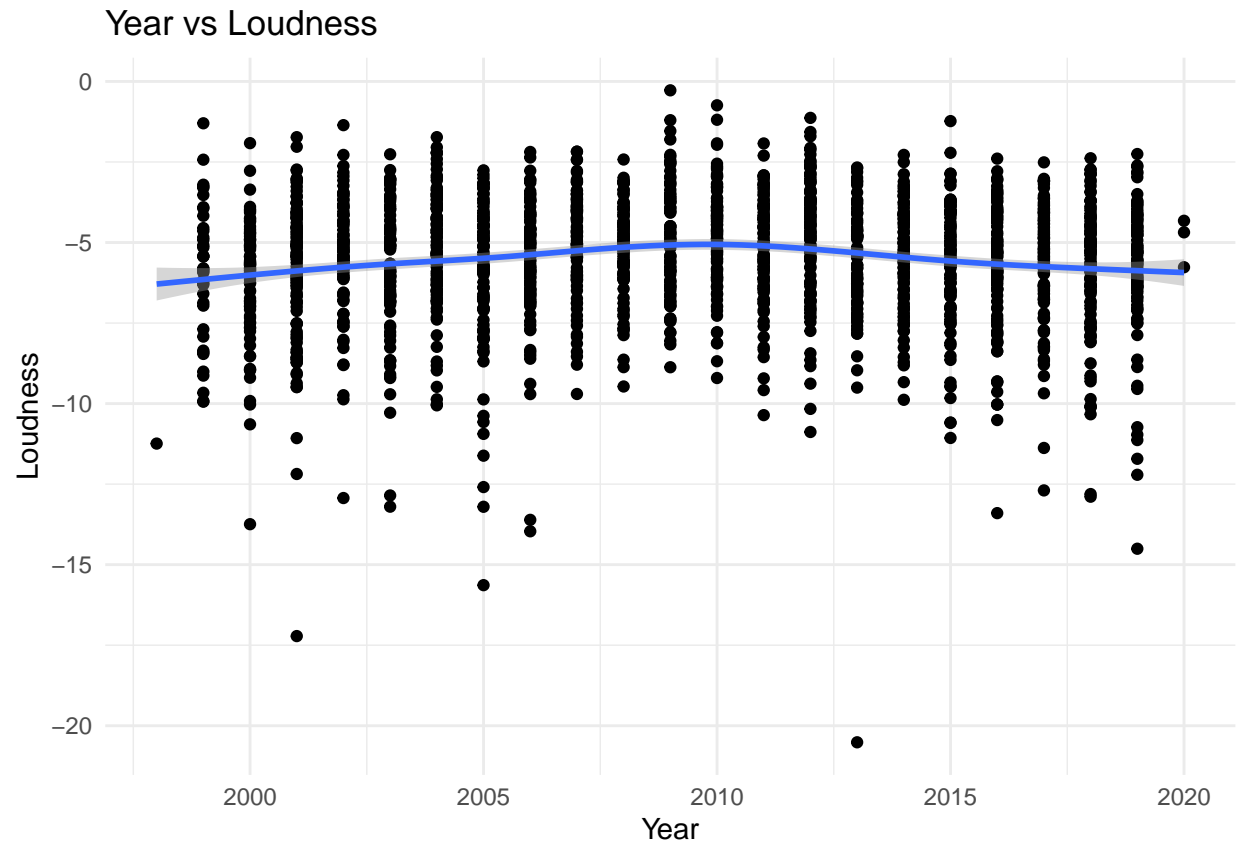
```
## [1] -9.974
```

```
q3 + iqr *1.5
```

```
## [1] -0.684
```

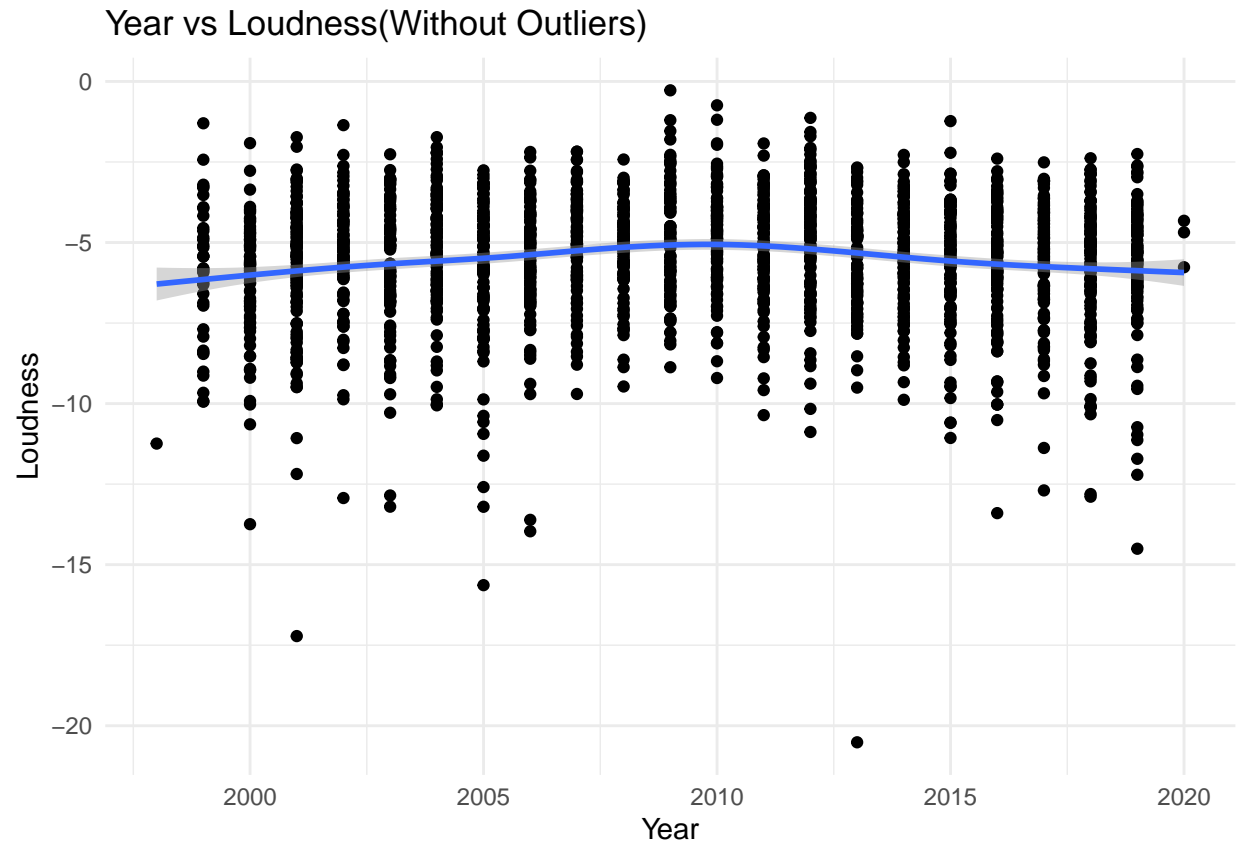
```
withoutOutliers <- filter(songs, loudness > 9.974, loudness < -0.684)
ggplot(data = songs, aes(x = year, y = loudness)) + geom_point() + theme_minimal() + labs(x = "Year", y = "loudness")
```

```
## 'geom_smooth()' using method = 'gam' and formula 'y ~ s(x, bs = "cs")'
```



```
ggplot(data = songs, aes(x = year, y = loudness)) + geom_point() + theme_minimal() + labs(x = "Year", y = "Loudness")

## 'geom_smooth()' using method = 'gam' and formula 'y ~ s(x, bs = "cs")'
```



```
unique(songs$genre)
```

```
## [1] "pop"
## [2] "rock, pop"
## [3] "pop, country"
## [4] "rock, metal"
## [5] "hip hop, pop, R&B"
## [6] "hip hop"
## [7] "pop, rock"
## [8] "pop, R&B"
## [9] "Dance/Electronic"
## [10] "pop, Dance/Electronic"
## [11] "rock, Folk/Acoustic, easy listening"
## [12] "metal"
## [13] "hip hop, pop"
## [14] "R&B"
## [15] "pop, latin"
## [16] "Folk/Acoustic, rock"
## [17] "pop, easy listening, Dance/Electronic"
## [18] "rock"
## [19] "rock, blues, latin"
## [20] "pop, rock, metal"
## [21] "rock, pop, metal"
## [22] "hip hop, R&B"
## [23] "pop, Folk/Acoustic"
```

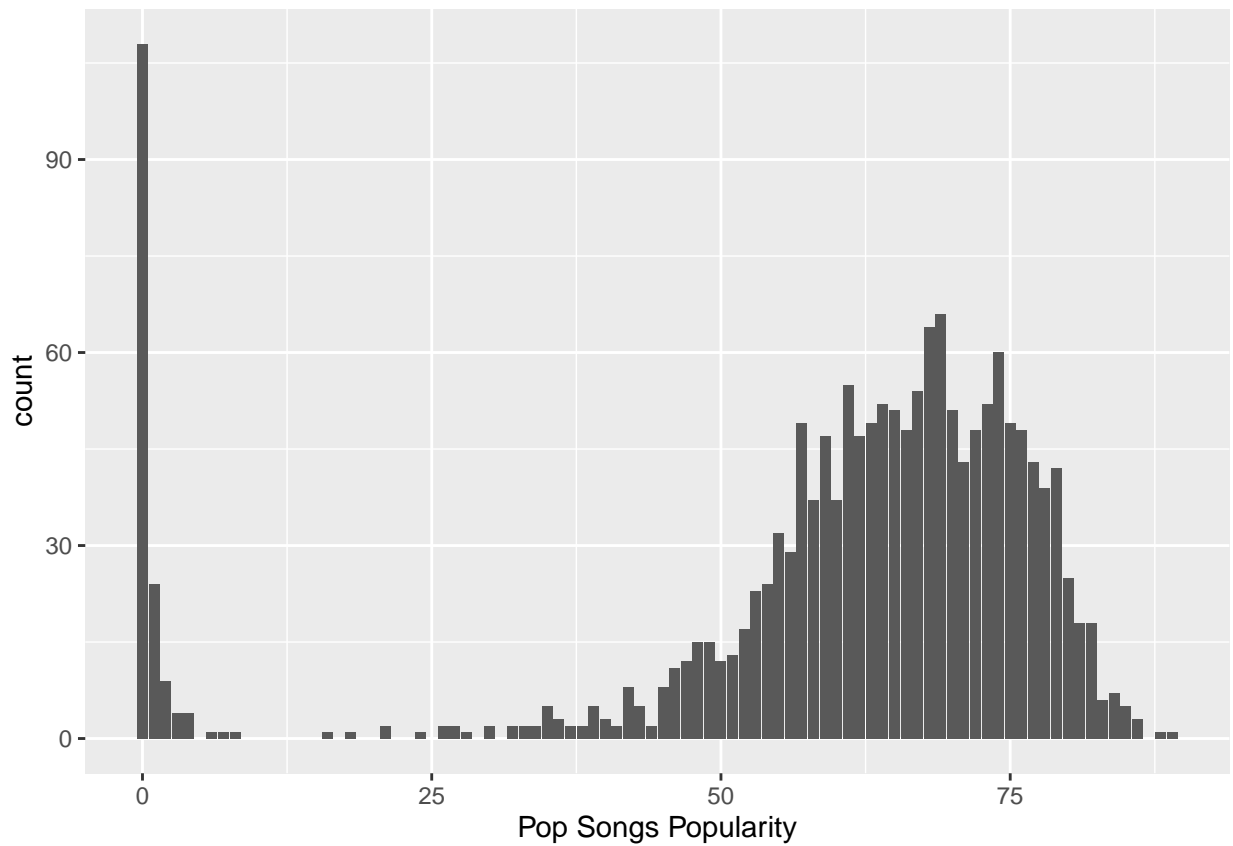
```
## [24] "set()"
## [25] "hip hop, pop, latin"
## [26] "hip hop, Dance/Electronic"
## [27] "hip hop, pop, rock"
## [28] "World/Traditional, Folk/Acoustic"
## [29] "Folk/Acoustic, pop"
## [30] "rock, easy listening"
## [31] "World/Traditional, hip hop"
## [32] "hip hop, pop, R&B, latin"
## [33] "rock, blues"
## [34] "rock, R&B, Folk/Acoustic, pop"
## [35] "latin"
## [36] "pop, R&B, Dance/Electronic"
## [37] "World/Traditional, rock"
## [38] "pop, rock, Dance/Electronic"
## [39] "pop, easy listening, jazz"
## [40] "rock, Dance/Electronic"
## [41] "World/Traditional, pop, Folk/Acoustic"
## [42] "country"
## [43] "hip hop, pop, Dance/Electronic"
## [44] "hip hop, pop, country"
## [45] "World/Traditional, rock, pop"
## [46] "World/Traditional, pop"
## [47] "hip hop, pop, R&B, Dance/Electronic"
## [48] "pop, R&B, easy listening"
## [49] "rock, pop, Dance/Electronic"
## [50] "Folk/Acoustic, rock, pop"
## [51] "rock, pop, metal, Dance/Electronic"
## [52] "pop, rock, Folk/Acoustic"
## [53] "country, latin"
## [54] "rock, classical"
## [55] "rock, Folk/Acoustic, pop"
## [56] "hip hop, rock, pop"
## [57] "easy listening"
## [58] "hip hop, latin, Dance/Electronic"
## [59] "hip hop, country"
```

```
popSongs <- filter(songs, str_detect(genre, "pop"))
#popSongs
hipHopSongs <- filter(songs, str_detect(genre, "hip hop"))
#hipHopSongs
rockSongs <- filter(songs, str_detect(genre, "rock"))
#hipHopSongs
metalSongs <- filter(songs, str_detect(genre, "metal"))
#metalSongs
bluesSongs <- filter(songs, str_detect(genre, "blues"))
#bluesSongs
edmSongs <- filter(songs, str_detect(genre, "Dance/Electronic"))
#edmSongs
countrySongs <- filter(songs, str_detect(genre, "country"))
#countrySongs
folkSongs <- filter(songs, str_detect(genre, "Folk/Acoustic"))
#folkSongs
latinSongs <- filter(songs, str_detect(genre, "latin"))
```

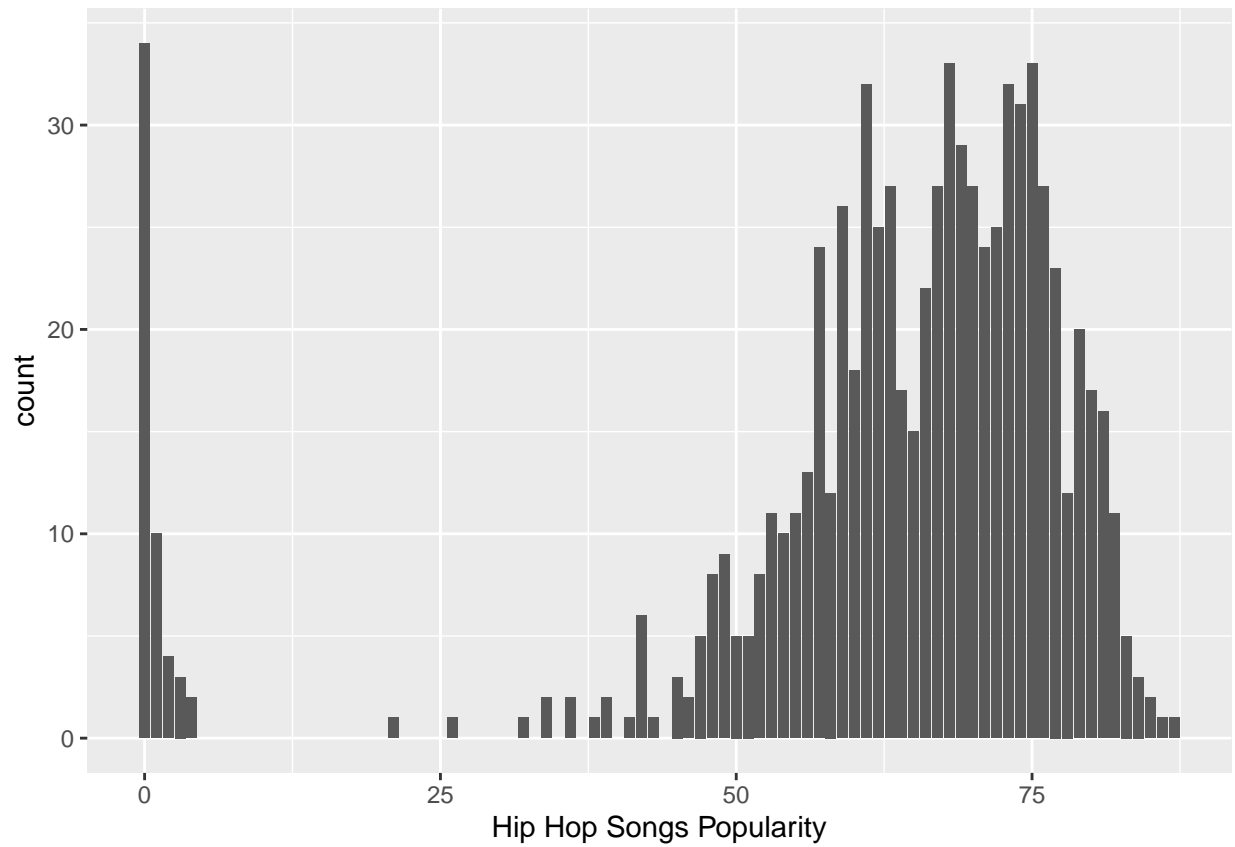


```
#latinSongs
RandBSongs <- filter(songs, str_detect(genre, "R&B"))
#RandBSongs
```

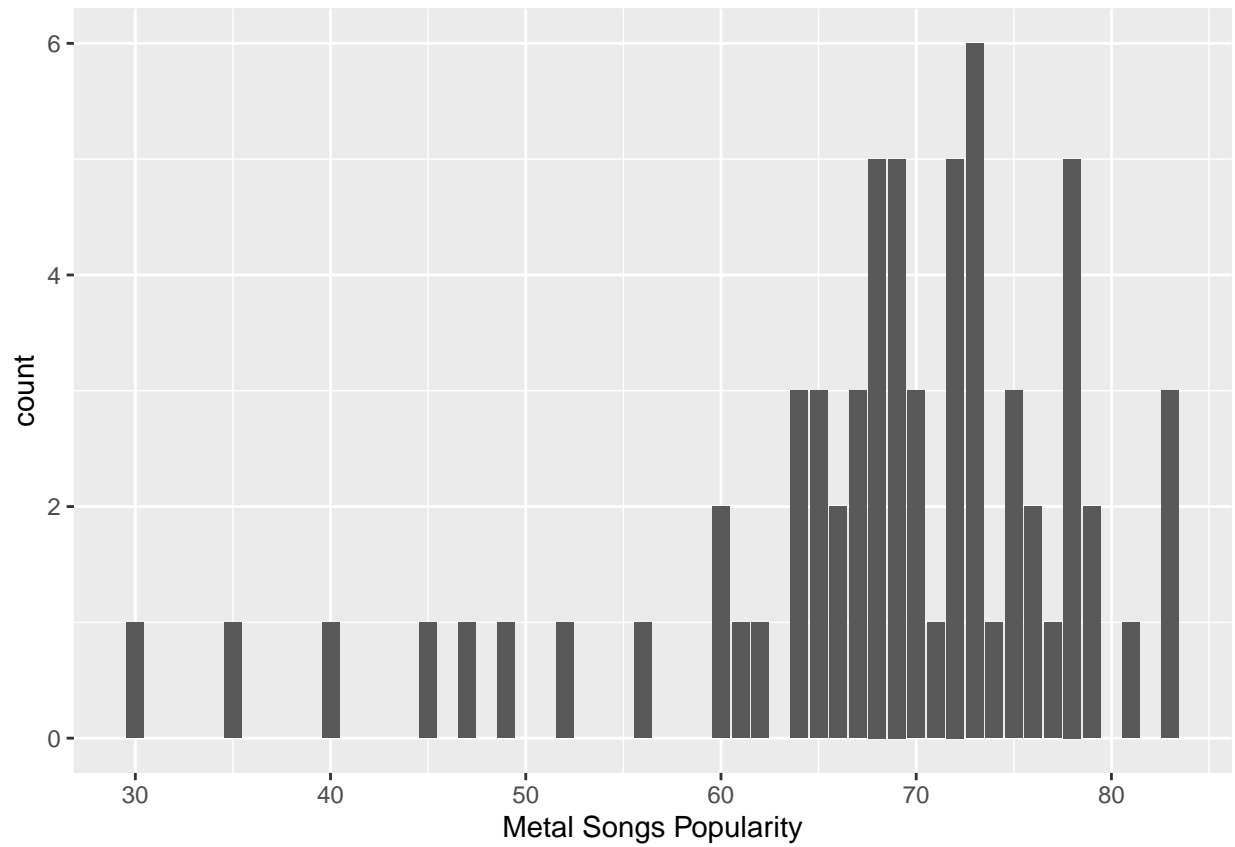
```
genres <- c(popSongs, hipHopSongs,hipHopSongs,metalSongs,bluesSongs,edmSongs ,countrySongs,folkSongs,la
ggplot(data = popSongs, aes(x = popularity)) + geom_bar() + labs(x = "Pop Songs Popularity")
```



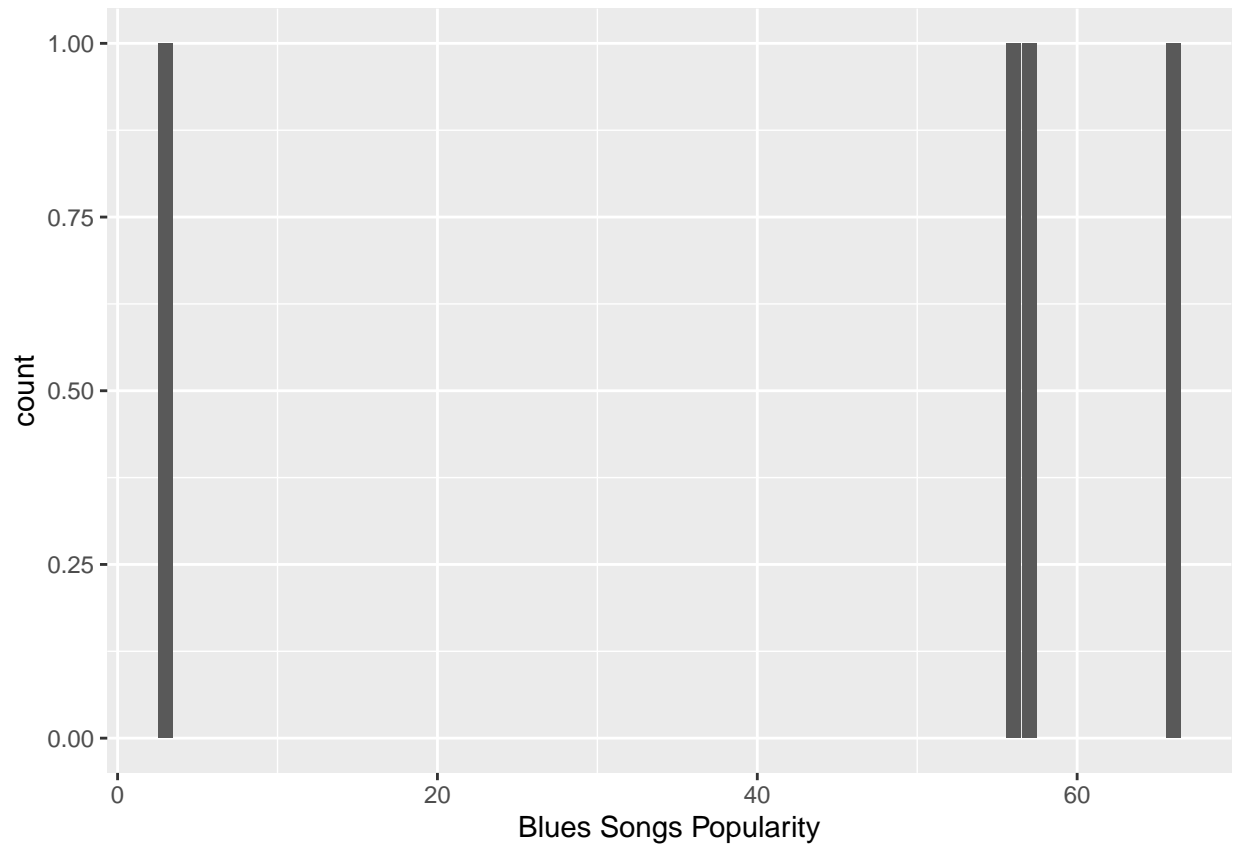
```
ggplot(data = hipHopSongs, aes(x = popularity)) + geom_bar() + labs(x = "Hip Hop Songs Popularity")
```



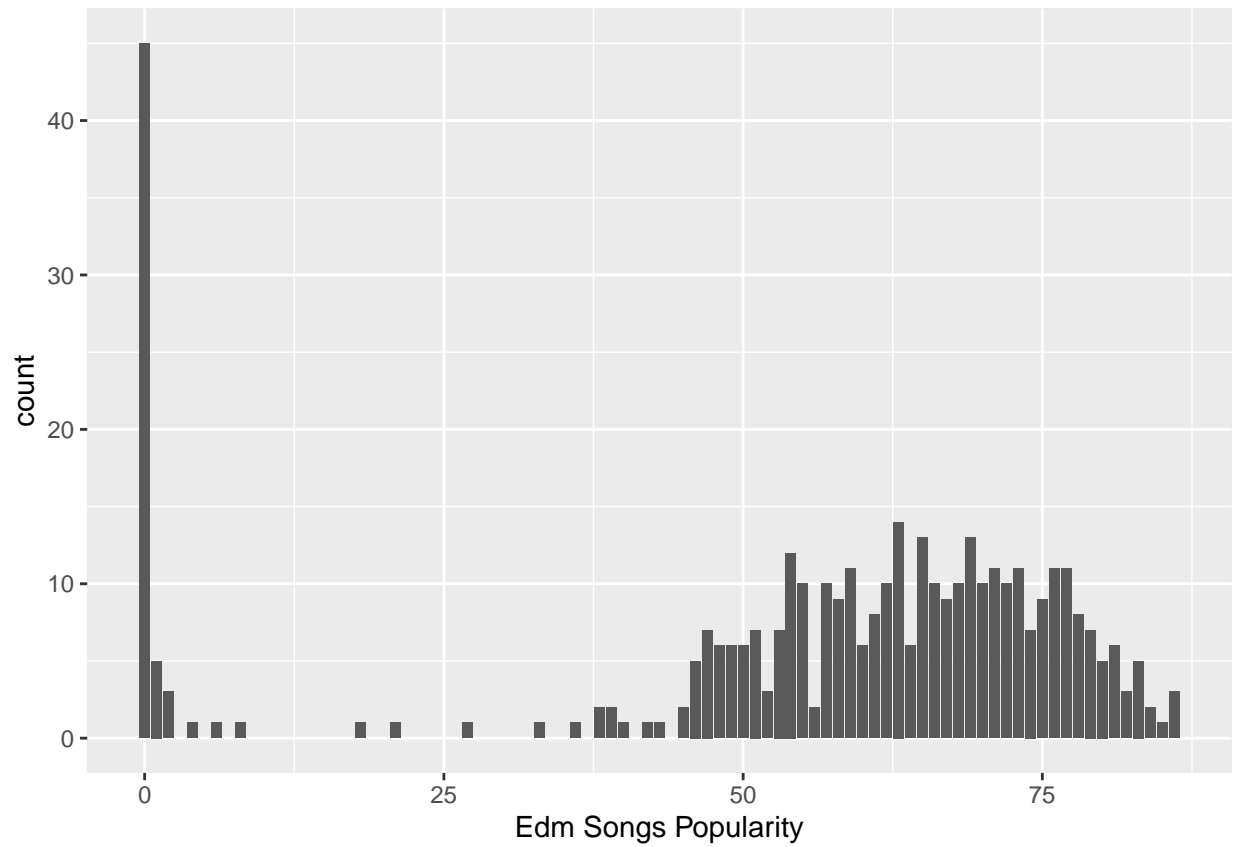
```
ggplot(data = metalSongs, aes(x = popularity)) + geom_bar() + labs(x = "Metal Songs Popularity")
```



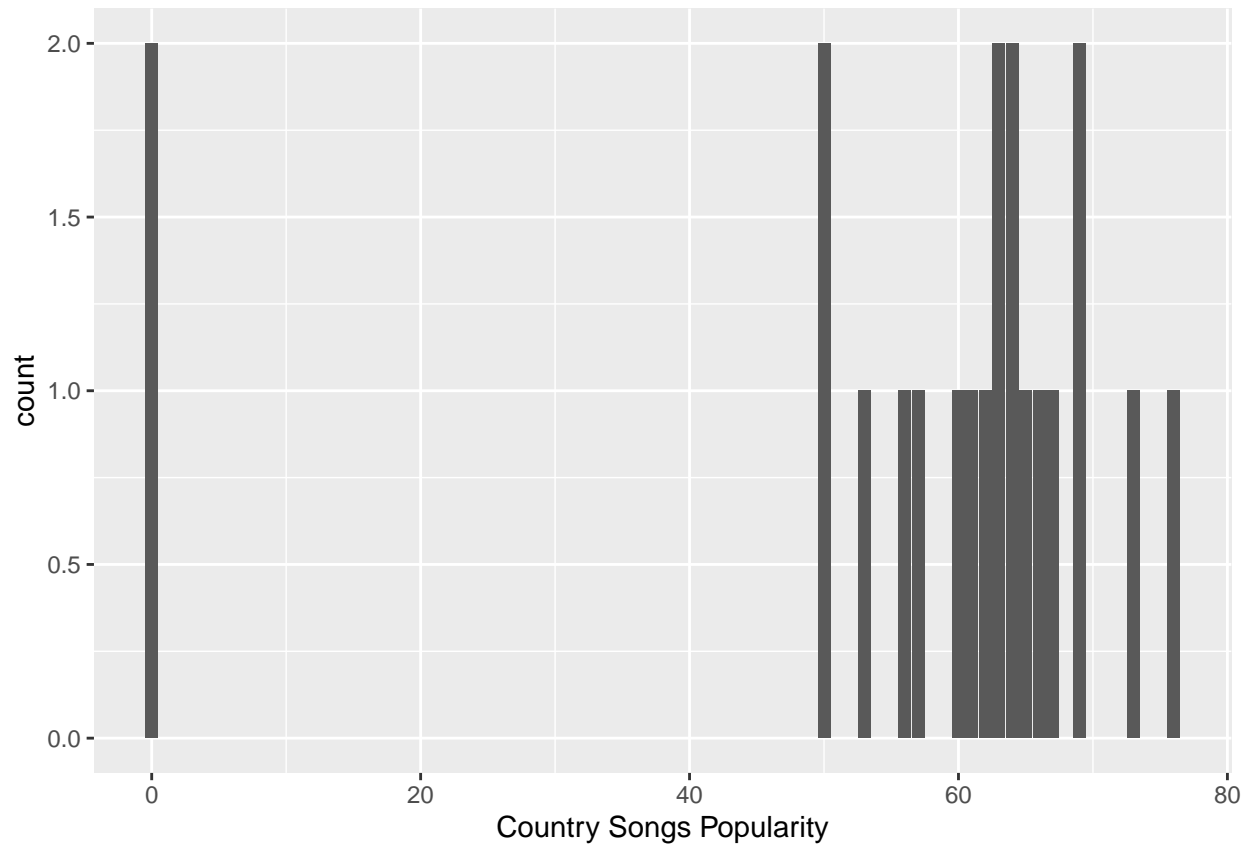
```
ggplot(data = bluesSongs, aes(x = popularity)) + geom_bar() + labs(x = "Blues Songs Popularity")
```



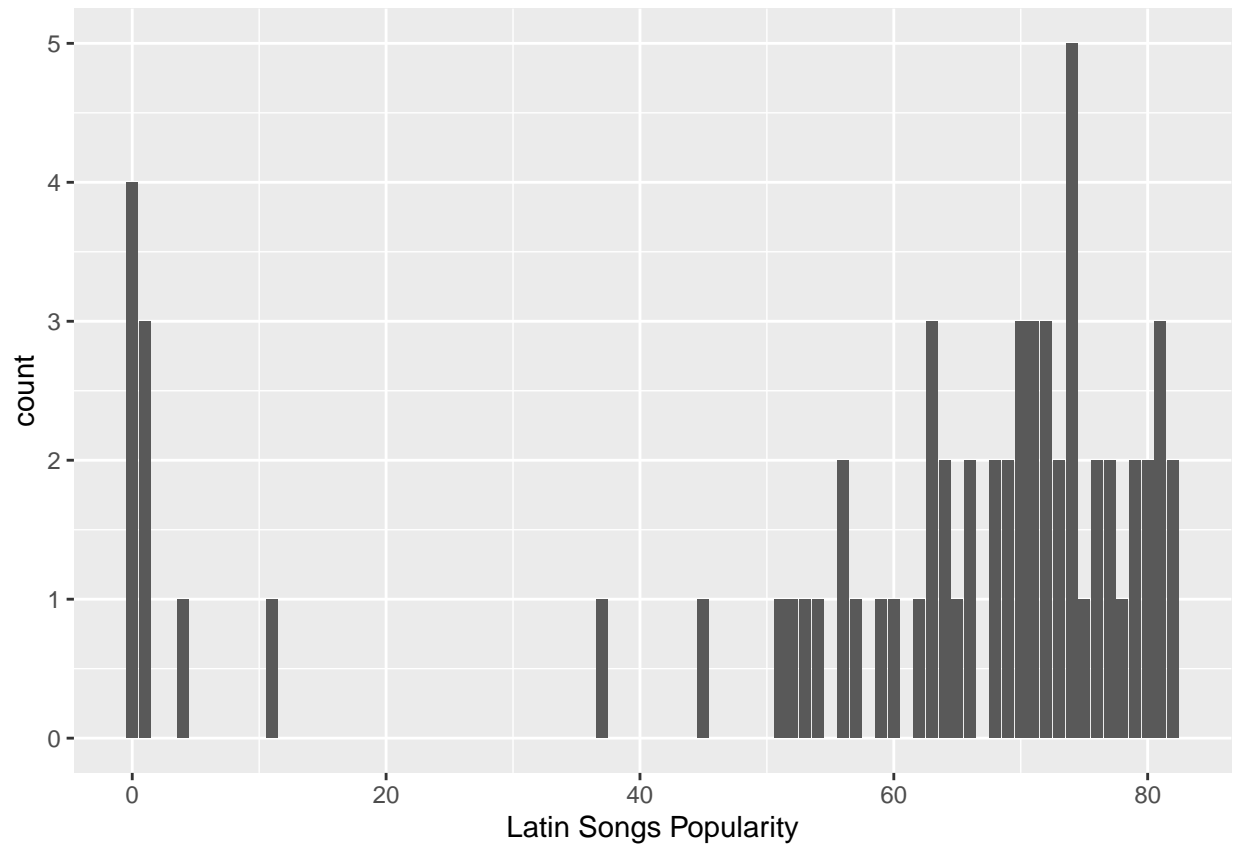
```
ggplot(data = edmSongs, aes(x = popularity)) + geom_bar() + labs(x = "Edm Songs Popularity")
```



```
ggplot(data = countrySongs, aes(x = popularity)) + geom_bar() + labs(x = "Country Songs Popularity")
```



```
ggplot(data = latinSongs, aes(x = popularity)) + geom_bar() + labs(x = "Latin Songs Popularity")
```



```
ggplot(data = RandBSongs, aes(x = popularity)) + geom_bar() + labs(x = "R&B Songs Popularity")
```

