

# Project report

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## Article Review

This article examines the recognition of hit 90's songs today in order to see how 90's music will be characterized by future generations. The article gathered data through a music quiz where participants were asked if they recognized certain Top 100 Billboard songs from the 90's. Different charts were produced to determine how songs stand the test of time and which songs are most recognized today.

## Description of the Dataset

Using a markdown table describe the variables from your dataset that you used for your analysis and whether they were in the original table or were obtained by transforming the original dataset. For variables that were not in the original dataset, briefly describe how you obtain them.

Variable Name	Description
artist_song	Artist name and song title
generation	Number of years until birth of subject when the song was released
recognition	Proportion of subjects that recognized the song
latest.recognition	Latest recognition data point (songs were debuted in different years, this is the last data point for a song)
diff.from.trend	The difference between the song's popularity and the average popularity for that age

The first three variables were from the original datasets.

## Helper functions used later

```
flip_sign <- function(l) {  
  l <- as.numeric(l)*(-1)  
  parse(text=l)  
}  
  
last_non_na_value <- function(r) {  
  cols <- length(r)  
  col <- cols - 1  
  while (is.na(r[col])) {  
    col <- col - 1  
  }  
  r[col]  
}
```

## Prepare Dataset

```
x <- nrow(rec)  
while (x > 0) {  
  y <- 25  
  while (y > 1) {
```

```

    if (rec[x, y] == 0) {
      rec[x, y] <- NA
      y <- y - 1
    } else {
      break
    }
  }
  x <- x - 1
}

```

## Main Conclusions

Use well-written paragraphs and professional-looking tables and/or graphs to reproduce the main results of your article of interest.

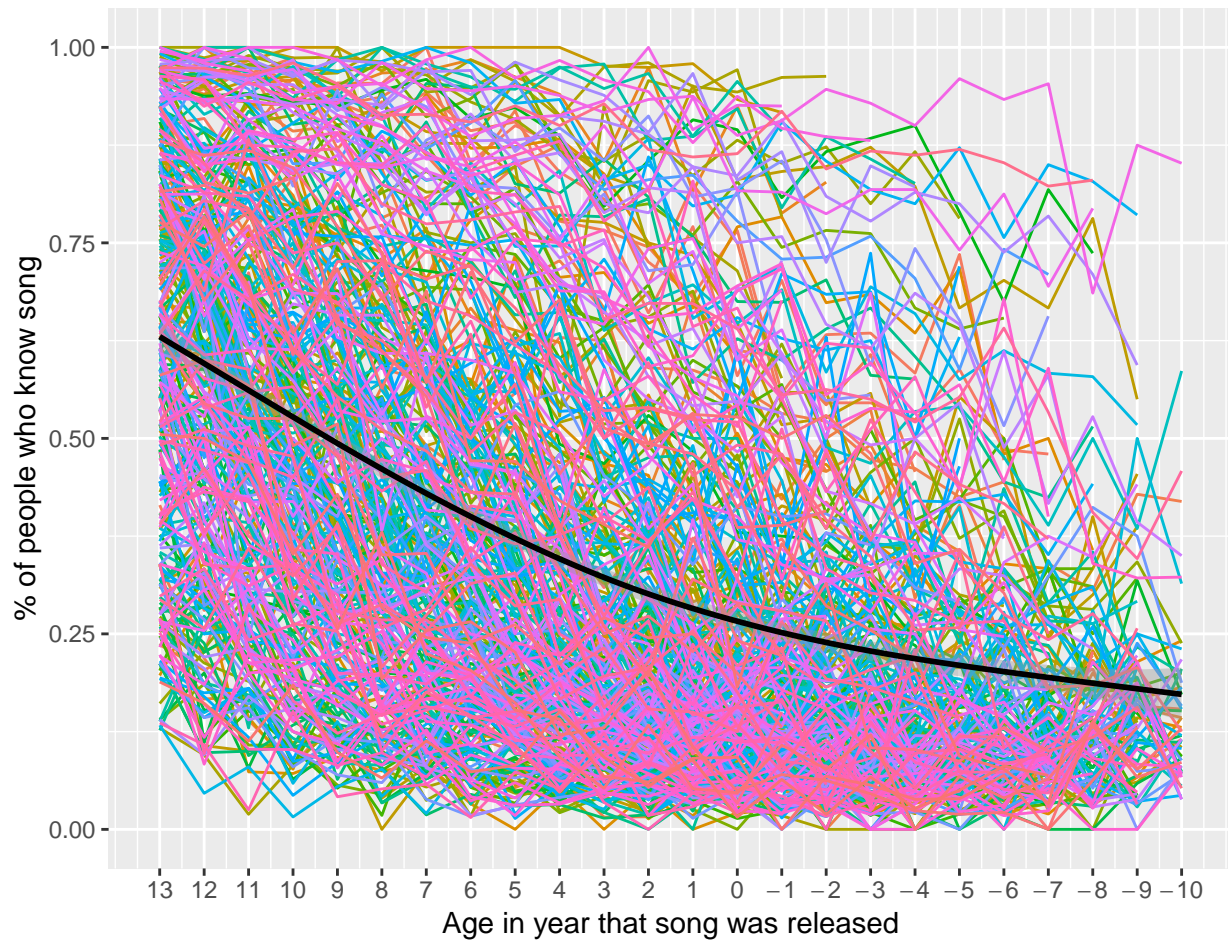
The percent of songs from the 90's recognized currently decays with time

```

rec %>%
  pivot_longer(`-13`:`10`,
               names_to = "generation", values_to = "recognition") %>%
  transmute(song = artist_song, generation = generation, recognition = recognition) %>%
  filter(!is.na(recognition)) %>%
  ggplot() +
  geom_line(aes(x = as.numeric(generation), recognition, color = song)) +
  geom_smooth(aes(as.numeric(generation), recognition), color = "black") +
  scale_x_continuous(breaks = seq(-13, 10, by = 1), labels = flip_sign) +
  theme(legend.position = "none") +
  labs(x = "Age in year that song was released",
       y = "% of people who know song")

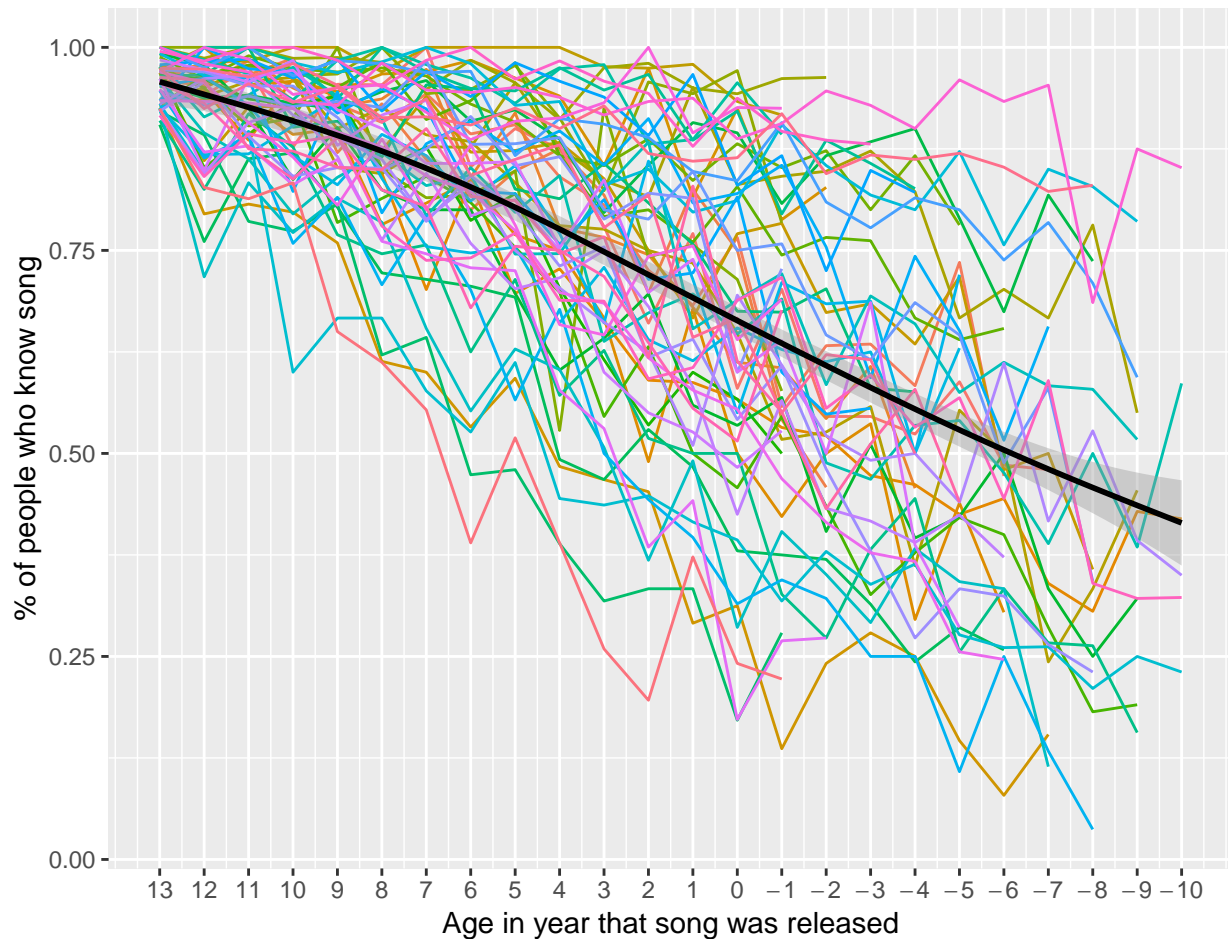
## `geom_smooth()` using method = 'gam' and formula 'y ~ s(x, bs = "cs")'

```



```
rec %>%
  filter(`-13` >= .9) %>%
  pivot_longer(`-13`:`10`,
               names_to = "generation", values_to = "recognition") %>%
  transmute(song = artist_song, generation = generation, recognition = recognition) %>%
  filter(!is.na(recognition)) %>%
  ggplot() +
  geom_line(aes(x = as.numeric(generation), recognition, color = song)) +
  geom_smooth(aes(as.numeric(generation), recognition), color = "black") +
  scale_x_continuous(breaks = seq(-13, 10, by = 1), labels = flip_sign) +
  theme(legend.position = "none") +
  labs(x = "Age in year that song was released",
       y = "% of people who know song")
```

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



A graph of the top-5 / bottom-5 songs as they are currently remembered

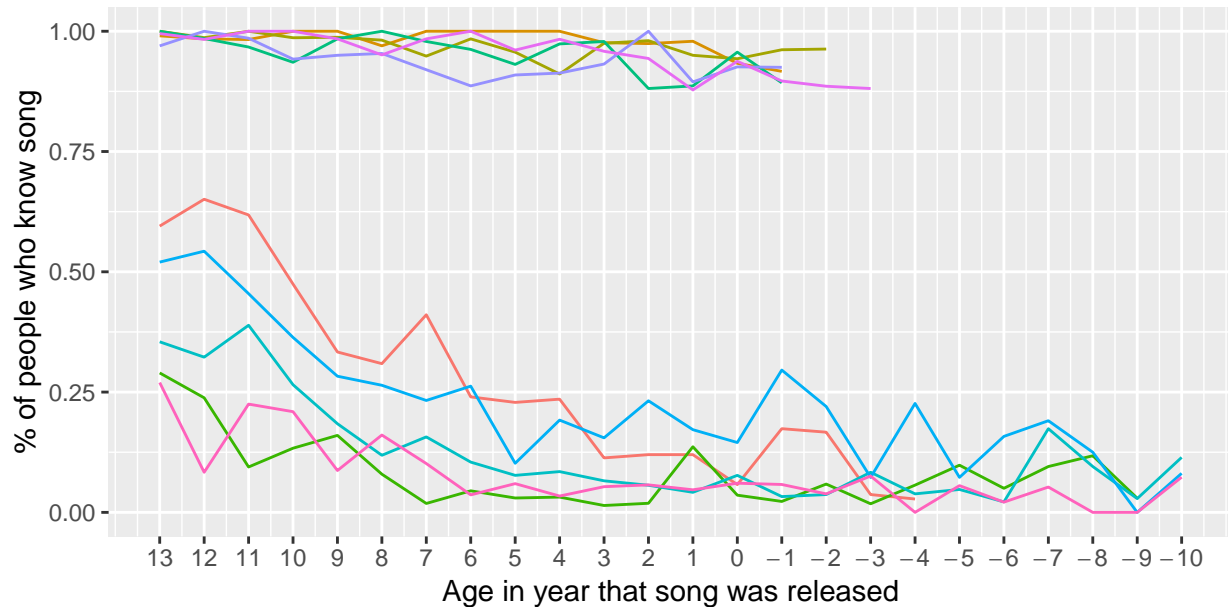
```
current.rec <- rec %>%
  rowwise() %>%
  transmute(artist_song = artist_song,
            latest.recognition = last_non_na_value(c(`-13`, `-12`, `-11`, `-10`, `-9`,
            `-8`, `-7`, `-6`, `-5`, `-4`,
            `-3`, `-2`, `-1`, `0`, `1`,
            `2`, `3`, `4`, `5`, `6`,
            `7`, `8`, `9`, `10`))) %>%

  arrange(desc(latest.recognition))

top.bottom.5.songs <- union(head(current.rec, 5), tail(current.rec, 5)) %>%
  pull(artist_song)

rec %>%
  filter(artist_song %in% top.bottom.5.songs) %>%
  pivot_longer(`-13`:`10`,
               names_to = "generation", values_to = "recognition") %>%
  transmute(song = artist_song, generation = generation, recognition = recognition) %>%
  filter(!is.na(recognition)) %>%
  ggplot() +
  geom_line(aes(x = as.numeric(generation), recognition, color = song)) +
```

```
scale_x_continuous(breaks = seq(-13, 10, by = 1), labels = flip_sign) +
theme(legend.position = "bottom", legend.direction = "vertical") +
guides(color = guide_legend(ncol=2)) +
labs(x = "Age in year that song was released",
     y = "% of people who know song",
     color = "Artist/Song")
```



#### Artist/Song

Bone Thugs N Harmony   Tha Crossroads	M.C. Hammer   Have You Seen Her
Britney Spears   .Baby One More Time	Poison   Unskinny Bop
Celine Dion   My Heart Will Go On	Smash Mouth   All Star
Jesus Jones   Real, Real, Real	Spice Girls, The   Wannabe
Lou Bega   Mambo No. 5	Taylor Dayne   Love Will Lead You Back

A graph of the top-5 songs that were popular in the 90's but are rarely heard after that

```
avg.table <- summarize_at(rec,
  vars(`-13`:`10`),
  ~ mean(., na.rm = TRUE))
```

```
avgs <- c()
avgs["-13"] = avg.table$`-13`
avgs["-12"] = avg.table$`-12`
avgs["-11"] = avg.table$`-11`
avgs["-10"] = avg.table$`-10`
avgs["-9"] = avg.table$`-9`
avgs["-8"] = avg.table$`-8`
avgs["-7"] = avg.table$`-7`
avgs["-6"] = avg.table$`-6`
avgs["-5"] = avg.table$`-5`
```

```

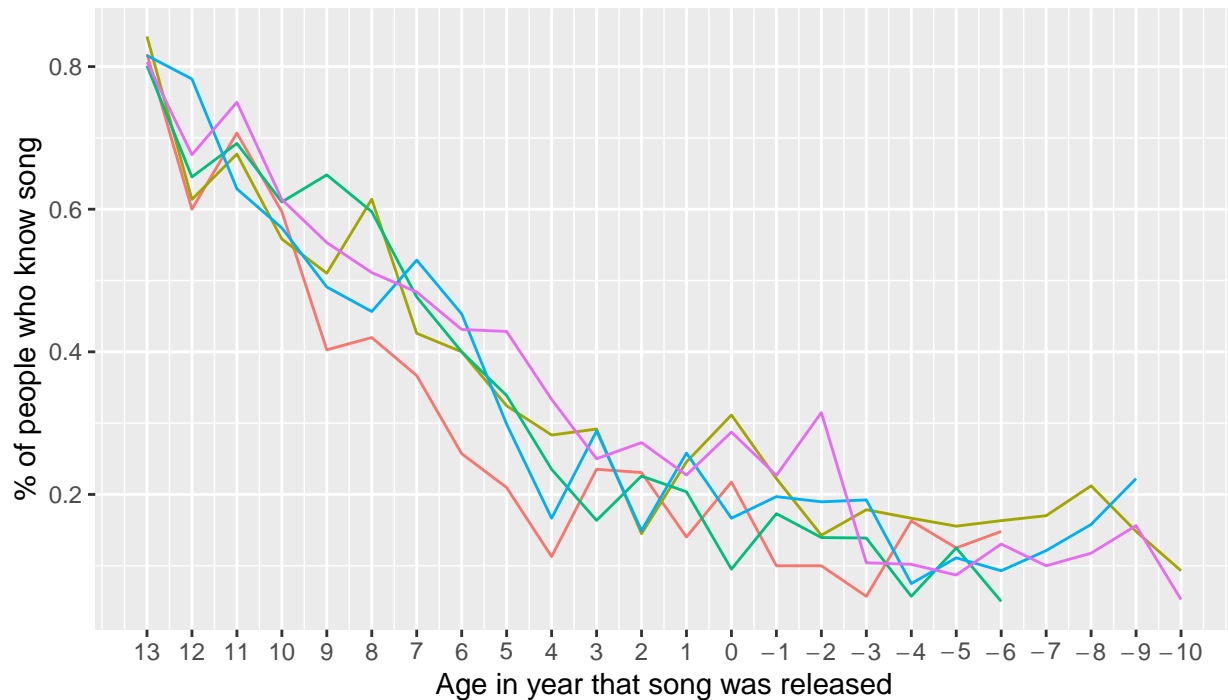
avgs["-4"] = avg.table$`-4`
avgs["-3"] = avg.table$`-3`
avgs["-2"] = avg.table$`-2`
avgs["-1"] = avg.table$`-1`
avgs["0"] = avg.table$`0`
avgs["1"] = avg.table$`1`
avgs["2"] = avg.table$`2`
avgs["3"] = avg.table$`3`
avgs["4"] = avg.table$`4`
avgs["5"] = avg.table$`5`
avgs["6"] = avg.table$`6`
avgs["7"] = avg.table$`7`
avgs["8"] = avg.table$`8`
avgs["9"] = avg.table$`9`
avgs["10"] = avg.table$`10`

low_songs <- rec %>%
  filter(`-13` >= .8) %>%
  pivot_longer(`-13`:`10`,
               names_to = "generation", values_to = "recognition") %>%
  transmute(song = artist_song, generation = generation, recognition = recognition) %>%
  filter(!is.na(recognition)) %>%
  mutate(diff = recognition - avgs[as.character(generation)]) %>%
  group_by(song) %>%
  summarize(diff.from.trend = sum(diff)) %>%
  arrange(desc(diff.from.trend)) %>%
  tail(5) %>%
  pull(song)

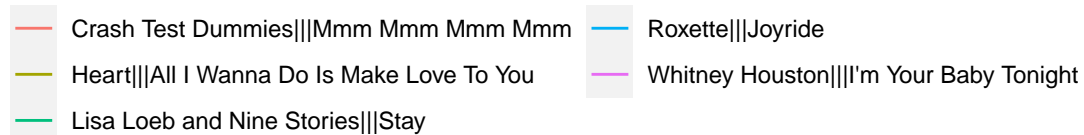
rec %>%
  filter(artist_song %in% low_songs) %>%
  pivot_longer(`-13`:`10`,
               names_to = "generation", values_to = "recognition") %>%
  transmute(song = artist_song, generation = generation, recognition = recognition) %>%
  filter(!is.na(recognition)) %>%
  ggplot() +
  geom_line(aes(x = as.numeric(generation), recognition, color = song)) +
  scale_x_continuous(breaks = seq(-13, 10, by = 1), labels = flip_sign) +
  theme(legend.position = "bottom", legend.direction = "vertical") +
  guides(color = guide_legend(ncol=2)) +
  labs(x = "Age in year that song was released",
       y = "% of people who know song",
       color = "Artist/Song")

```





#### Artist/Song



A graph/table describing the change in recognition across different generations

```
knitr::kable(head(arrange(generation.rec.data,
                           desc(generation.rec.data$mean_gen_z_recognition)), 50),
  digits = 2,
  col.names = c("Artist/Song", "Millenial", "Gen Z"))
```

Artist/Song	Millenial	Gen Z
Celine Dion   My Heart Will Go On	0.97	0.96
Britney Spears   ..Baby One More Time	0.99	0.96
Spice Girls, The   Wannabe	0.98	0.93
Smash Mouth   All Star	0.95	0.92
Lou Bega   Mambo No. 5	0.98	0.92
Cher   Believe	0.98	0.90
Snap!   The Power	0.92	0.88
Los Del Rio   Macarena	0.97	0.87
Backstreet Boys, The   Everybody	0.97	0.86
Whitney Houston   I Will Always Love You	0.91	0.85
R. Kelly   I Believe I Can Fly	0.96	0.82
Coolio   Gangsta's Paradise	0.92	0.82
Michael Jackson   Black Or White	0.89	0.81
House Of Pain   Jump Around	0.89	0.80
Ricky Martin   Livin' La Vida Loca	0.92	0.79

Artist/Song	Millenial	Gen Z
Sir Mix-A-Lot   Baby Got Back	0.84	0.77
R.E.M.   Losing My Religion	0.89	0.74
Elton John   Can You Feel The Love Tonight	0.91	0.72
Peabo Bryson and Regina Bell   A Whole New World	0.86	0.70
Christina Aguilera   Genie In A Bottle	0.92	0.69
C and C Music Factory   Gonna Make You Sweat	0.87	0.69
TLC   No Scrubs	0.89	0.67
Aerosmith   I Don't Want To Miss A Thing	0.91	0.67
Joan Osborne   One Of Us	0.91	0.66
Sixpence None The Richer   Kiss Me	0.89	0.66
Puff Daddy and Faith Evans   I'll Be Missing You	0.89	0.65
Proclaimers, The   I'm Gonna Be	0.87	0.65
Michael Jackson   You Are Not Alone	0.77	0.63
Montell Jordan   This Is How We Do It	0.85	0.62
Marky Mark and The Funky Bunch   Good Vibrations	0.79	0.62
Enrique Iglesias   Bailamos	0.91	0.60
Ace Of Base   The Sign	0.88	0.60
Alanis Morissette   Ironie	0.89	0.59
Toni Braxton   Un-Break My Heart	0.90	0.58
Santana   Smooth	0.93	0.57
Kris Kross   Jump	0.72	0.55
Red Hot Chili Peppers   Under The Bridge	0.84	0.54
Ace Of Base   All That She Wants	0.83	0.54
Billy Ray Cyrus   Achy Breaky Heart	0.80	0.53
Celine Dion   All By Myself	0.75	0.53
BLACKstreet   No Diggity	0.79	0.53
Seal   Kiss From A Rose	0.88	0.51
Tag Team   Whoomp! There It Is	0.82	0.51
TLC   Waterfalls	0.85	0.50
Nicki French   Total Eclipse Of The Heart	0.68	0.47
Madonna   Vogue	0.69	0.47
Vanilla Ice   Ice Ice Baby	0.68	0.46
Third Eye Blind   Semi-Charmed Life	0.73	0.44
Real McCoy   Another Night	0.75	0.44
Notorious B.I.G., The   Hypnotize	0.63	0.44

## Original question

We want to know how recognized songs are today that were only recognized by less than 50% of people that were 13-15 years old at the time of release. Do some of these songs gain popularity or do most die out?

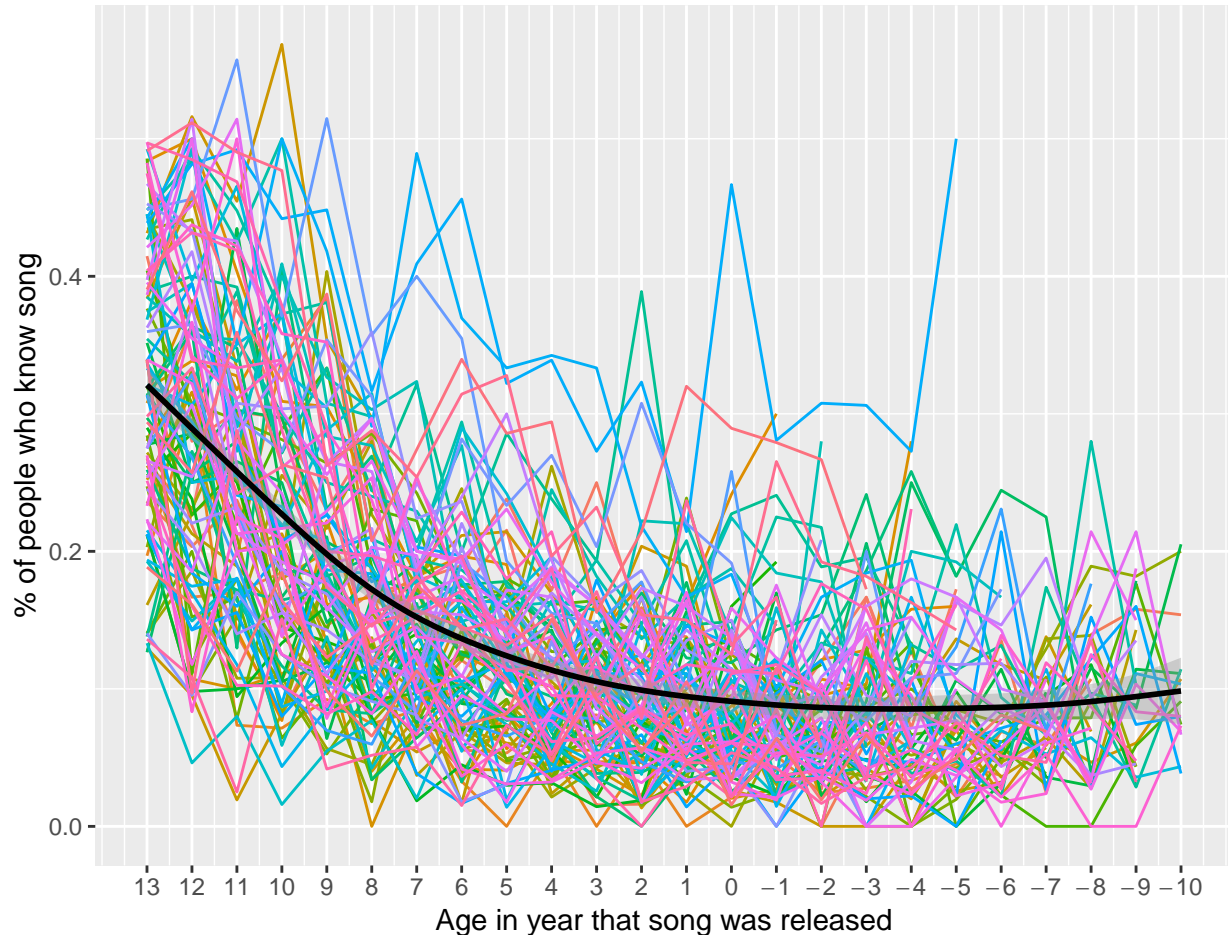
```
low_rec_at_release <- rec %>%
  filter(`-13` <= .5) %>%
  pivot_longer(`-13`:`10`,
               names_to = "generation", values_to = "recognition") %>%
  transmute(song = artist_song, generation = generation, recognition = recognition) %>%
  filter(!is.na(recognition))

ggplot(low_rec_at_release) +
  geom_line(aes(x = as.numeric(generation), recognition, color = song)) +
  geom_smooth(aes(as.numeric(generation), recognition), color = "black") +
  scale_x_continuous(breaks = seq(-13, 10, by = 1), labels = flip_sign) +
```



```
theme(legend.position = "none") +
labs(x = "Age in year that song was released",
     y = "% of people who know song")
```

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



```
low_rec_at_release %>%
  mutate(diff = recognition - avgs[as.character(generation)]) %>%
  group_by(song) %>%
  summarize(diff.from.trend = sum(diff)) %>%
  arrange(desc(diff.from.trend)) %>%
  head(5) %>%
  pull(song)
```

```
## [1] "Notorious B.I.G., The|||The What"
## [2] "Notorious B.I.G., The|||One More Chance-Stay With Me"
## [3] "Whitney Houston|||I Believe In You And Me"
## [4] "Whitney Houston|||Exhale"
## [5] "Brandy|||Have You Ever?"
```

Most of these songs appear to die out and not be recognized today. We can tell this by the decreasing slope of the trend line for these songs that we recognized by less than 50% of 13-15 year olds when the song was released. This isn't too surprising of a result. It is interesting to see which of these songs are most recognized today though. The top 2 songs are by Notorious B.I.G. and the next 2 most recognized are by Whitney Houston. It makes sense that these artists are at the top since they are still popular today and have many

fans that could recognize some of their least popular songs.