

AE 05: Mathematical Models

Songs on Spotify

Driver: _____, Reporter: _____, Gopher: _____

2024-09-11

! Important

- Open [RStudio](#) and create a subfolder in your AE folder called “AE-05”
- Go to the [Canvas](#) and locate your AE 05 assignment to get started.
- Upload the `ae-05.qmd` and `spotify-popular.csv` files into the folder you just created. The `.qmd` and PDF responses are due in Canvas no later than Saturday, September 14 at 11:59pm.

```
library(tidyverse)
library(ggformula)
library(broom)
library(knitr)
library(patchwork) #arrange plots in a grid
```

Data

The data set for this assignment is a subset from the [Spotify Songs](#) Tidy Tuesday data set. The data were originally obtained from Spotify using the **spotifyr** R package.

It contains numerous characteristics for each song. You can see the full list of variables and definitions [here](#). This analysis will focus specifically on the following variables:

variable	class	description
<code>track_id</code>	character	Song unique ID
<code>track_name</code>	character	Song Name
<code>track_artist</code>	character	Song Artist
<code>track_popularity</code>	double	Song Popularity (0-100) where higher is better

variable	class	description
energy	double	Energy is a measure from 0.0 to 1.0 and represents a perceptual measure of intensity and activity. Typically, energetic tracks feel fast, loud, and noisy. For example, death metal has high energy, while a Bach prelude scores low on the scale. Perceptual features contributing to this attribute include dynamic range, perceived loudness, timbre, onset rate, and general entropy.
valence	double	A measure from 0.0 to 1.0 describing the musical positiveness conveyed by a track. Tracks with high valence sound more positive (e.g. happy, cheerful, euphoric), while tracks with low valence sound more negative (e.g. sad, depressed, angry).

```
spotify <- read_csv("spotify-popular.csv")
```

Are high energy songs more positive? To answer this question, we'll analyze data on some of the most popular songs on Spotify, i.e. those with `track_popularity` ≥ 80 . We'll use linear regression to fit a model to predict a song's positiveness (`valence`) based on its energy level (`energy`).

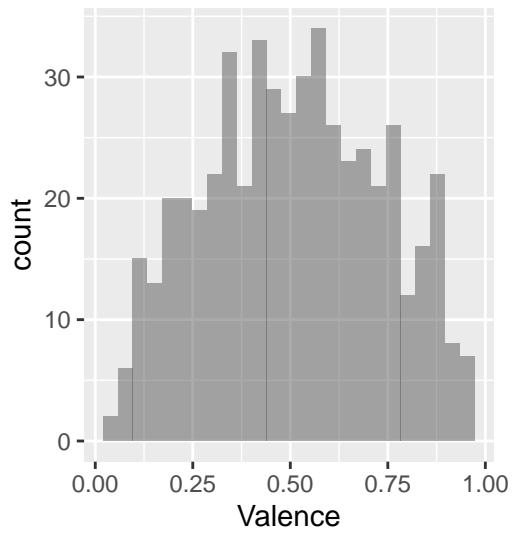
Below are plots as part of the exploratory data analysis.

```
p1 <- gf_histogram(~valence, data = spotify) |>
  gf_labs(title = "Distribution of Valence",
    subtitle = " for Popular songs on Spotify",
    x = "Valence")

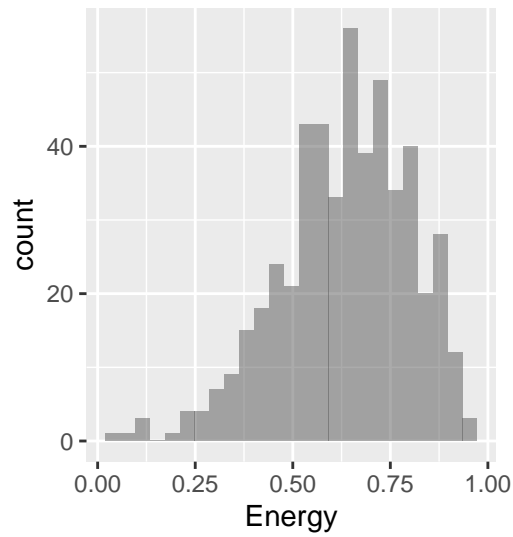
p2 <- gf_histogram(~energy, data = spotify) |>
  gf_labs(title = "Distribution of Energy",
    subtitle = "for Popular songs on Spotify",
    x = "Energy")

p1 + p2 # The patchwork package will arrange your plots for you
```

Distribution of Valence
for Popular songs on Spotify

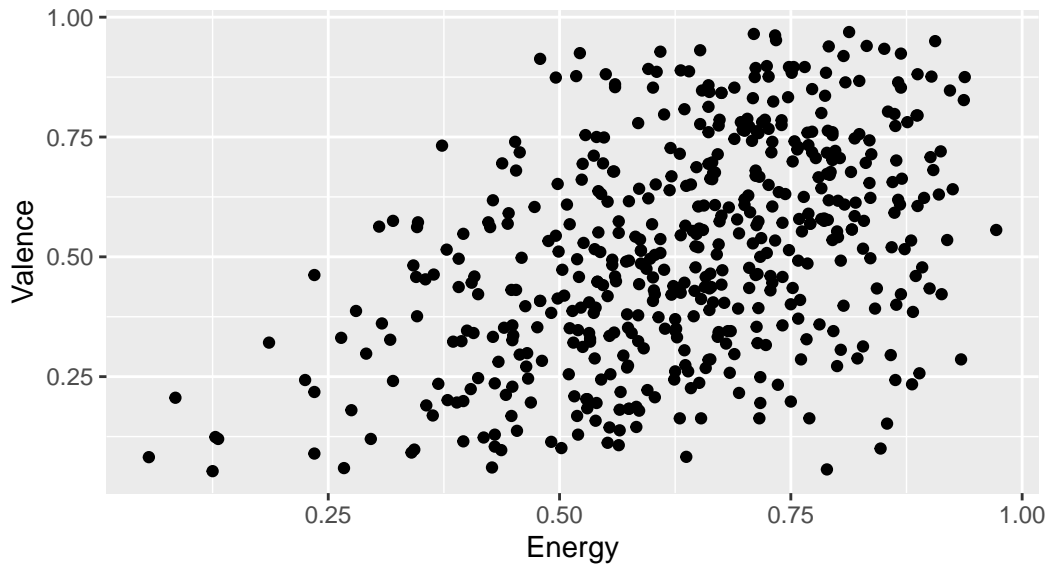


Distribution of Energy
for Popular songs on Spotify



```
gf_point(valence ~ energy, data = spotify) |>
  gf_labs(title = "Valence vs. Energy",
    subtitle = "Popular songs on Spotify",
    x = "Energy",
    y = "Valence")
```

Valence vs. Energy
Popular songs on Spotify



Exercise 1

Fit a model using the **energy** of a song to predict its **valence**, i.e. positiveness. Include the 90% confidence interval for the coefficients, and display the output using 3 digits.

```
## add code
```

Exercise 2

In words interpret the estimate and confidence interval for the slope in the previous exercise.

Exercise 3

Interpret the p-value from Exercise 1.

Exercise 4

Predict what the average valence for a song with an energy score 0.5 is. Report and interpret a 90% confidence interval for the average valence.

Exercise 5

Report and interpret a 90% confidence interval for a single song with energy score 0.8.

! Important

To submit the AE:

- Render the document to produce the PDF with all of your work from today's class.
- Upload your qmd and pdf files to the Canvas assignment.