# Predicting Song Composition with Spotify Data Central London Data Science

Wifi:welovecode

# Collecting data

#### Research datasets

- Well curated, great for prototyping or benchmarking
- Already labelled
- Too sterile- May not be representative of real life

### Curate your own:

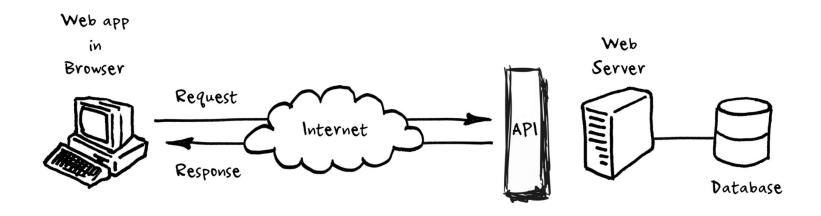
- Arduous
- Expensive

# Collecting data

Why not get someone else to do it?

- Companies are collecting as much data as they can
- Significant amount of effort in curating it for you
- It will be real world events

# **Application Programming Interfaces**



## Types of music

What is a genre?

- Hiphop, Metal, Acid jazz...
- Is it a group of people or culture that surround a type of music and label it accordingly?
- Or is it music that sounds distinct to any other type of music?

Let's just stick with the latter tonight....

## **Timbre**

Distinguishes one sound from another

Same amplitude, same frequency same note != same timbre

We use words like:

- Sharp
- Eerie
- Bright

# **Spotify**

- Music streaming service
- Access to millions of songs
- Super rich API



# Starting with this...

{'meta': {'analyzer\_version': '4.0.0', 'platform': 'Linux', 'detailed\_status': 'OK', 'status\_co de': 0, 'timestamp': 1444833490, 'analysis\_time': 21.35246, 'input\_process': 'libvorbisfile L+R 44100->22050'}, 'track': {'num\_samples': 11539500, 'duration': 523.33333, 'sample\_md5': '', 'of fset\_seconds': 0, 'window\_seconds': 0, 'analysis\_sample\_rate': 22050, 'analysis\_channels': 1, 'end\_of\_fade\_in': 0.0, 'start\_of\_fade\_out': 523.33331, 'loudness': -6.82, 'tempo': 125.152, 'te mpo\_confidence': 0.078, 'time\_signature': 4, 'time\_signature\_confidence': 0.877, 'key': 11, 'ke y\_confidence': 0.738, 'mode': 1, 'mode\_confidence': 0.664, 'codestring': 'eJxVm1eC4zgSRK-iI8Cb-19s3wuwZmZ\_uouQBIJpIiMN9xzlrF7Wr\_xqLaevc9tv1PVr9bZVVuv8NcZvtjXabfXXWym\_ede6tc7-G5MPx6xj137Wb7VR vOztzHV\_7Dx-e5yy5-jnt085v9ZW3Vy132mT-\_DPGrv23y3nfJetzt\_te\_3G8r79118tc5XfmfXc0SvXdZ7xu23v3Su\_rpy reH3GbrVx3Wf78Qh3tjsO14N\_OUc95R6etrOh18tb8HmfJfutfeq8XJ\_G2c\_kifZh\_zHbyf3KKY3fj8v53v2qv5-tc9fFXf pl5zqH\_5bZ776z-IXNsXy0woPz1Xn7QpK7IK7NDXhuJMu9xhmda56r5bqXiiw5d833W0HAXvPXD9Hes3bbv3rrOsp-8QiNz -\_wAQpSvWdy\_-v91H47vSCaUlU1560cwWvvv0\_j1-Vsrs9dv3PmRqmbz-viAf19Z8fh9eb3fH\_Mw134BMHwL8bEP9W\_esyo 79Iqmu2FXRBYnSiB3\_f5j4Iaz9-wvSiw3qWAm0d7Al4qqI2zBgJFzGvnhhOZY11zHA4xuN5oiBueU\_q9vxbreze8le8vLYh jFoRa-HwXH3jxa02fawwhFnDZY3Ktwj4No5C27zqfRWhR7egn7DV70-La-VNgL93PLzb5H4W12znlvWtzroKAuclSYWfVMu 4PNWu3e87ZsXuuJwJGeh1Vr8X14X5cTzWzfx2z0781-HHdS89Es7-uubNH51oFDu411ujz13sbcbDTc3LUE7XpcP34-UY-s 3H\_dSf7jSh0rn3LYrvxxLmuHuPHe37iWVqa\_v0PA-KaHTMY0thBPMNrc0DHk\_Dc1Rv3XT8Hvus09t9zne\_79XKcfcSDeUu7 6KUfb394srM4ItdqcwxM5XDQX79FJysDb8Nz-MJtsZe1eK4FYuCW-i04VerpXmtwQNkeVcn3OMqSHTBQ9o5DADYFDwZhRlV \_6LNy5Da51qGew865ueYG7qe9IaCB3k\_0L5byeRs-x-oc1Z0MpqPjQKhZsBwNw-W6FzxzoLleY29ct55rxXw5BnDRdF31DV 4jqbNB4eF5bwFYwFnuryR\_uMrEJmZquZVn332BcGN-iBf7FsVvzvNn39hq-S\_iDfyi\_QfxxsaA\_qNw70Eo8Mn1EGJHAJ4vI

# Turning it into something that looks like this...

