# INVOLUNTARY MUSICAL IMAGERY 

## -INVESTIGATING MUSICAL FEATURES THAT PREDICT 'EARWORIIS'-

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## "Roadmep"

- Introduction
- Terminology
- Previous findings
- The idea behind this study
- Met̂́nods
- Fiesulis
- Conclusion anad Discussion


## "Roadmep"

- Intirocluction
- Methods
- How to find genuine earworms
- How to analyze InMI tunes
- Fiesulis
- Conclusion anad Discussion


## "Roadmap"

- Introcluction
- Metinods
- Results
- The earworm formula!?
- Conclusion arricl Disciussion


## "Roadmap"

- Introcluction
- Methods
- Resulis
- Conclusion and Discussion
- How to interpret the found features
- How to shape future research


## Terminology

- Involuntary musical imagery (InMI)
- Likkanen (2008)
- Song in Your Head Phenomenon
- Spontaneously, Repeatedly, Involuntarily
- Earworm
- Derived from 'Ohrwurm' (German)
- Levitin (2006), Sacks (2007), BBC 6 Radio


## Previous finclings

- Liikkanen (2008)
- 90\% experience earworms daily
- Only 15\% describe them disturbing
- Beaman \& Williams (in press)
- Earworm episode less than 24 hours
- Earworm itself longer than short term memory capacity would suggest
- Hemming (2008)
- Importance of genre and lyrics


## The idea behind this study

- No study has dealt with musical features of earworms yet.
- Are earworms different?
- De la Motte (1993)
- Analyzed his personal earworms:
- repetitive motif, harmonically appealing, only 3-5 tones
- Müllensiefen \& Kopiez (in press)
- Musical features can predict success of cover songs


## How to find genuine earworms

- Online Survey
- 1014 participants
- 35.6 years (SD= 13.4 years; range 13-76 years)
- 572 females and 441 males
- Recent earworm <-> Frequent earworms
- Artist, song title, exact part
- 1449 usable earworm tracks
- Top earworm list -> 75 songs (6\%)
- Named more than once
- In total: 227 (16\%)
- 14.000 files MIDI Corpus


## Top 5 earvorris

| artist | song | incs |
| :--- | :--- | :---: |
| Lady Gaga | Bad romance | 13 |
| Journey | Don't stop believing | 11 |
| Lady Gaga | Alejandro | 11 |
| Katy Perry | California gurls | 10 |
| Kylie Minogue | Can't get you out of my <br> head | 7 |

## How to find genuine earworms

- Using UK chart data to control for:
- Popularity (exposure)
- Recency effects
- 52 songs left
- Predictors
- hi.entry: Highest chart position
- exit.date: Days from end of study to last chart appearance
- weeks: Number of weeks in the charts
- entry.date: Days from end of study to first chart appearance
- Response
- incs: Number of namings



## How to find genuine earworms

## Poisson Model



Wald's Chi-square test:
$X^{2}(2, N=110)=19.218, p<0.001^{* * *}$

## How to find genuine earworms

- Positive residual deviance
$\square$ More often named than expected from the model
- Named more than once
$\square$ More likely to be genuine
- 29 earworms


## How to emelyze Irivy turnes

- Findings matching non-earworms
- Random draw from MIDI corpus
- 150 (UK chart data available)
- Not named as earworms
- Gower's Dissimilarity coefficient


## Gower's Dissimilerity

- Measuring similarity between two objects, using numeric and character variables
- We are using:
- hi.entry
- entry.date
- exit.date
- weeks
- genre
- artist
- Matrix -> lowest value for each earworm

You never gonna get this song....

## How to ennelyze Iriyly turnes

- 29 earworm tracks

| artist | song | incs | hi.entry | weeks | entry.date | exit.date | genre |
| :--- | :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| lady gaga | bad romance | 13 | 1 | 38 | 281 | 15 | pop |
| lady gaga | alejandro | 11 | 7 | 10 | 253 | 183 | pop |
| journey | don't stop <br> believing | 11 | 6 | 47 | 477 | 149 | rock |
| katy perry | californiagurls | 10 | 1 | 6 | 43 | 1 | pop |
| queen | bohemian <br> rhapsody | 7 | 1 | 17 | 12699 | 12580 | rock |

- 29 non-earworm tracks

| artist | song | incs | hi.entry | weeks | entry.date | exit.date | genre |
| :--- | :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| gorillaz | feel good inc. | 0 | 2 | 39 | 1940 | 1667 | pop |
| jessica <br> these boots are <br> made for <br> walkin' |  | 0 | 4 | 10 | 1800 | 1730 | pop |
| handbags and | 0 | 4 | 15 | 3164 | 3059 | rock |  |
| stereophonics | gladrags |  |  |  |  |  |  |
| my place | 0 | 1 | 11 | 2164 | 2087 | pop |  |
| nelly | 0 | 1 | 13 | 12054 | 11963 | rock |  |
| elvis presley | way down | 0 |  |  |  |  |  |

## Modelling

## Logistic Regression

- Predictor variables:
- 40 musical features
- 12 clusters
- Response variable
- Binary earworm status
- (1 = yes, 0 = no)


## Step AIC

- Stepwise algorithm for model selection
- Using Akaike information creterion
- Simplifying the the logistic regression


## Resuilts

- Logistic regression model:
- Using 4 features

|  | Estimate | Std. Error | $\mathbf{z}$ value | $\operatorname{Pr}(>\|\mathbf{z}\|)$ |
| :---: | :---: | :---: | :---: | :---: |
| (Intercept) | -7.7520 | 4.1703 | 0.9386 | 0.0630 . |
| d.median | 0.0767 | 0.0373 | 2.0613 | $0.0393^{*}$ |
| tonal.clarity | 5.9946 | 3.4817 | 1.7218 | 0.0851. |
| int.cont.grad.std | -0.3878 | 0.1989 | -1.9597 | 0.0512. |
| i.leaps | 41.8001 | 20.3481 | 2.0543 | 0.0399 * |

- Predicts $72 \%$ of the data set correctly
$-X^{2}(4, N=58)=8.7476, p=.0677$

Model M1


## How to interpret the features

- d.median
- the median of the average duration of all notes
- int.cont.grad.std
- standard deviation of interpolation contour measure
- tonal.clarity
- how clear is the tonality of the melody
- Auhagen (1994)
- i.leaps
- average number of leaps larger than a $5^{\text {th }}$
- Rauhe (1987) "Activation structures"


## Conclusions

- Songs that appear often as earworms can be distinguished from other pop songs
- Model predicts $72 \%$ correctly
- Using only musical features
- Excluding contextual \& subject-related variables


## How to shepe further research

- Better ways to control for exposure
- Airplay charts, API queries (lastfm)
- Hurdle and negative binomial models
- Increasing number of possible matches
- Different earworm types?
- Decision tree models
- Corpus features
- Including context and subject-related variables


## Asking bigger cuuestions

- Have we found the ultimate pop song formula?
- Are successful songs earworm OR earworms commercially bestselling?
- Can we learn something about musical memory?
- Müllensiefen \& Halpern (submitted)
- Musical features predict implicit and explicit memory for melodies


## How to shepe further research

Project is ongoing!!!
Any ideas are welcome!

## Thank you for your attention



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