ALGORITHMIC MUSIC COMPOSITION

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HISTORY

Mozart's dice game

"Anleitung zum Componieren von Walzern so viele man will vermittelst zweier Würfel, ohne etwas von der Musik oder Composition zu verstehen"



=> 45,949,729,863,572,161 different yet similar waltzes

- 1955 Hiller, Isaacson: First computer-generated composition
- **1991** Gibson, Byrne: Musical Composition Using Genetic Algorithms And Neural Networks
- **2011** Donnelly, Sheppard: Evolving Four-Part Harmony Using Genetic Algorithms

MUSIC COMPOSITION AS SEARCH PROBLEM

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Goal:

 $\cdot\,$ Compose music without or with minimal human guidance

Approach:

- \cdot Search the set of all possible compositions
- · Return one which sounds good

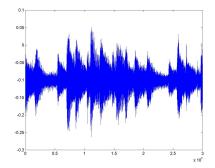
Problems:

- · How to represent music?
- · Search space is very big how to search efficiently?
- · How to evaluate if something sounds good?

How can we represent music?

- $\cdot\,$ Audio files
- · Flat structure
- · Hierarchical structure

MUSIC REPRESENTATION

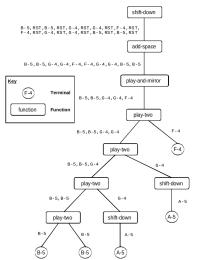


Source: http://www.angelfire.com/art2/speech-audio-seperat/

MUSIC REPRESENTATION



Source: http://www.well.com/user/bryan/last.gif



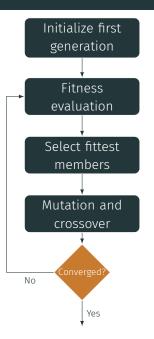
A-5, RST, A-5, RST, F-4, RST, F-4, RST, E-4, RST, E-4, RST, F-4, RST, F-4, RST, A-5, RST, A-5, RST

Source: http://graphics.stanford.edu/ bjohanso/papers/gp98/johanson98gpmusic.pdf

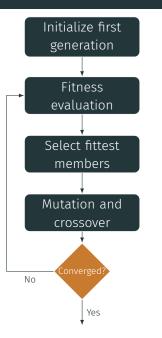
How can we search efficiently?

· Genetic algorithms

GENETIC ALGORITHMS



GENETIC ALGORITHMS



Example: Maximize function $f:\mathbb{R}^n\mapsto\mathbb{R}$

- \cdot Initialization: Select $x_1,\ldots x_k \in \mathbb{R}^n$ at random
- Fitness of x: f(x)
- · **Mutation:** Shift by random $\varepsilon \in \mathbb{R}^n$
- **Crossover** of x and y: Choose value from line-segment between x and y

How can we evaluate music?

- \cdot Human based
- \cdot Rule based
- · Machine learning based

On a scale from 1 to 10, how well did you like thi	s?
Cancel OK	



Source: https://en.wikipedia.org/wiki/Johann_Joseph_Fux



Source: http://conferences.telecom-bretagne.eu/fps2012/program/slides/07.pdf

EVOLVING FOUR-PART HARMONY ...

Patrick Donnelly, John Sheppard Evolving Four-Part Harmony Using Genetic Algorithms Applications of Evolutionary Computation, 2011, pp 273-282

Used techniques:

- · Flat music representation
- · Genetic algorithms
- · Rule based Fitness

MUSIC REPRESENTATION

Music representation:

- · Four parallel parts
- · Each part: List of (pitch, duration) tuples
- $\cdot\,$ In C-major
- $\cdot\,$ Total length not fixed

Example:

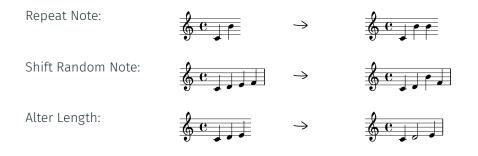


Genetic Algorithms

Initialize first generation with C-major chords



Mutation: Apply one of these operations to a random part



Total: 11 operations

Crossover: Cut and glue two elements together







Rule Based Fitness

What does good music need?

- · Melody (Should be catchy)
- · Harmony (Melodies should interact nicely)
- · Rhythm (Emphasize meter)
- · Structure (Intro, outro, reoccurring themes, ...)
- \cdot Timbre/Intonation

 \Rightarrow Music theory provides tools to enforce these constraints

- Leap Height: Two consecutive notes should not have an interval larger than a 9th
- Voice Crossing: An upper part should always play higher than a lower part
- **Opening/Closing Chord:** The piece should start and end with a C-major chord
- · Intervals: Pure and dissonant intervals should be avoided
- Total: 15 rules

 $fitness(rule_i) = \frac{n_i - v_i}{n_i}$

total fitness = $\sum_{i} (\omega_i \cdot \text{fitness}(\text{rule}_i))$

- $\cdot \, n_i$ = Number of places where rule i could be violated
- v_i = Number of places where rule i actually is violated

EXAMPLE

 $fitness(rule_i) = \frac{n_i - v_i}{n_i}$

- $\cdot \, n_i$ = Number of places where rule i could be violated
- $\cdot \, v_i$ = Number of places where rule i actually is violated



Leap Height:

- n_i = number of leaps = 10
- \cdot v_i = number of large leaps = 1
- fitness(Leap Hight) = $\frac{9}{10}$

Voice Crossing:

- \cdot n_i = num. of parallel notes = 4
- \cdot v_i = number of crossings = 2
- fitness(Voice Crossing) = $\frac{1}{2}$

Pro:

- · Strong music-theoretical foundation
- · Works well in practice

Contra:

- Humans need to define the rules for each genre (Cannot be automated)
- $\cdot\,$ Some genres are hard to express by rules

We discussed:

- · Genetic algorithms
- · Rule based fitness functions

Using these tools it is possible to generate interesting music without human interaction.

SOURCES

Papers:

Patrick Donnelly, John Sheppard
Evolving Four-Part Harmony Using Genetic Algorithms
Applications of Evolutionary Computation, 2011, pp 273-282

Images:

- https://en.wikipedia.org/wiki/Johann_Joseph_Fux
- http://conferences.telecombretagne.eu/fps2012/program/slides/07.pdf
- http://www.angelfire.com/art2/speech-audio-seperat/
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END