artificial intelligence and music

threats & opportunities

alexander lerch



intro

education

- Electrical Engineering (Technical University Berlin)
- Tonmeister (music production, University of Arts Berlin)

professional

- Associate Dean for Research & Creative Practice, College of Design, Georgia Tech
- Associate Professor, School of Music, Georgia Tech
- prev: 2000-2013: CEO at zplane.development

background

- machine learning for audio and music (20+ years)
- audio algorithm design (20+ years)
- commercial music software development (10+ years)
- entrepreneurship (10+ years)

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ww.linkedin.com/ir

introduction artificial intelligence

■ artificial intelligence

- unclear definition: everything that is perceived to act intelligently
- changes over time

■ machine learning

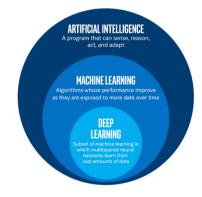
 data-driven: algorithm is more agnostic to task and is parametrized through training with data

■ deep learning

deep neural networks are the ML approach used

■ generative AI

• deep neural networks generating content





- mapping function is learned from patterns and characteristics from data
- ⇒ model success largely depends on training data
- technical challenges concerning data
 - imbalance & bias (distribution is skewed, biased)
 - diversity & representativeness
 - subjectivity of annotations
 - noisiness (bad quality, bad annotations, ...)
 - amount





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machine learning importance of data





machine learning: generic algorithm mapping an input to an output

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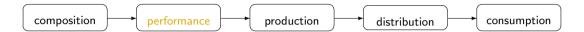
4 / 9

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- **creation of musical ideas** ("score")
 - defines style and idea
- realization of musical ideas into acoustical rendition
 - interpretation, modification, addition, and dismissal of score information
 - unique acoustic representation of score
- recording, mixing, and editing (in case of record media)
 - editing and splicing of recorded data: timbre, equalization choices
 - not separable from performance in a recording
- distribution & listening
 - music recommendation and discovery





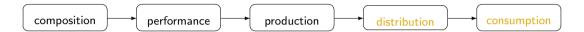
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musical communication musical communication and Al



composition

- intelligent assistance, e.g., ideas, auto-arrangements
- automatic composition
- performance
 - interactive music education systems
 - generation of 'human' performance
- production
 - auto-edit and auto-mix
- distribution
 - match music style and consumer
- consumption
 - intelligent music discovery & adaptable music



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music ai opportunities & threats

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opportunities

- content creation:
 - speed-up, increased efficiency
 - creative possibilities (morphing, etc.)
 - co-creative idea givers
 - democratization
- consumption:
 - personalization
 - effective discovery and accessibility

threats

- both:
 - 'mainstreamification'
 - bias through for-profit system control
 - sustainability and energy
- content creation:
 - ethical use of data
 - plagiarism growth
 - liability for harmful content
- consumption:
 - user distrust through
 - inflationary ai-generated content
 - ► inexplainable block-box systems

■ many opportunities

- increased efficiency in content production
- new tech will always be used in unforeseen creative ways
- accessibility increases dramatically

paradigm shift has to be actively managed

- management and mitigation of impact on workforce/livelihood
- transparency and informed consumers
- models for fair compensation

old questions worth asking anew

- when is a musical piece considered creative
- what makes a human performance unique
- can generated content be art



links

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book: www.AudioContentAnalysis.org

music informatics group: musicinformatics.gatech.edu





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