### Music AI: Threats and Opportunities

overview

alexander lerch



#### 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)



### introduction artificial intelligence

#### **■** artificial intelligence

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

#### **■** machine learning

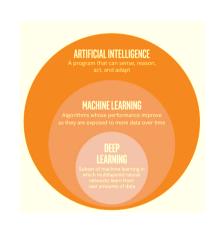
 data-driven: algorithm is more agnostic to task and is parametrized through training with data

#### **■** deep learning

deep neural networks are the ML approach used

#### ■ generative AI

• deep neural networks generating content





### machine learning: generic algorithm mapping an input to an output

- 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



# machine learning importance of data





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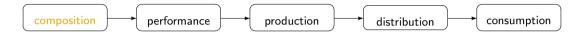




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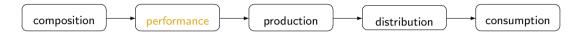
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- **creation of musical ideas** ("score")
  - defines style and idea
- realization of musical ideas into acoustical rendition
  - interpretation, modification, addition, and dismissal of score information
  - unique acoustic representation of score
- 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





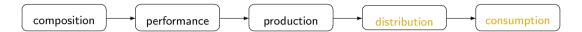
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### musical communication paradigm shift

- historical technological shifts in music:
  - recording devices (tape, grammophone)
  - digitization/softwarization of recording studio
- historical technological shifts in general:
  - internet
  - introduction of photography



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blues funk metal chanson<sup>a</sup>



ntro ai & ml music ai **evaluation** challenges conclusion thank

### systematic evaluation evaluation targets

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#### **■** system output

- originality
  - plagiarism
  - diversity
  - creativity
- audio quality
- musical & aesthetic qualities

#### user experience

#### other criteria

- explainability
- bias
- ethical use of data & data curation practices
- resource use & environmental impact



### systematic evaluation methods

#### ■ **subjective** testing

- preference test
- Turing test
- rating of properties

#### objective testing

- reference-independent
- comparison of distributions

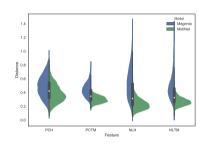
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### music ai opportunities & threats

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

- content creation:
  - ethical use of data
  - growth in plagiarism
  - liability for harmful content
  - livelihood of creators
  - value perception of artistic content
- consumption:
  - consumer distrust through
    - ▶ inflationary ai-generated content
    - ▶ inexplainable black-box systems
- **both**:
  - 'mainstreamification'
  - bias (data curation, for-profit system control)

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thanks

#### links

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music informatics group: musicinformatics.gatech.edu



