

# SleepGAN: Towards Personalized Sleep Therapy Music

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\* This work was done at Nokia Bell Labs during the author's internship



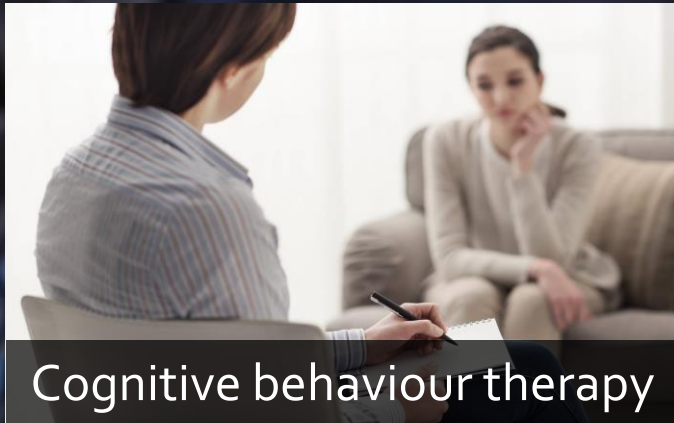




40% UK adults <sup>[1]</sup>



43% US adults <sup>[2]</sup>



Cognitive behaviour therapy



Pharmaceutical sleep aids

- Expensive
- Inconvenient
- Expert guidance
- Potential harmful side-effects

Alternative non-invasive, low-cost, and enjoyable solutions to sleep therapy

[1] Trahan et al., "The music that helps people sleep and the reasons they believe it works: A mixed methods analysis of online survey reports", 2018

[2] CDC declares sleep disorders a public health epidemic

# Music Therapy

Convenient and enjoyable

Proven effects in clinical studies

Little scalability (e.g. Oura ring app, 8 soundscapes)

Limits a mass adoption of sleep music apps

Music fatigue

Rarely consider individual music preference

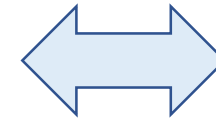
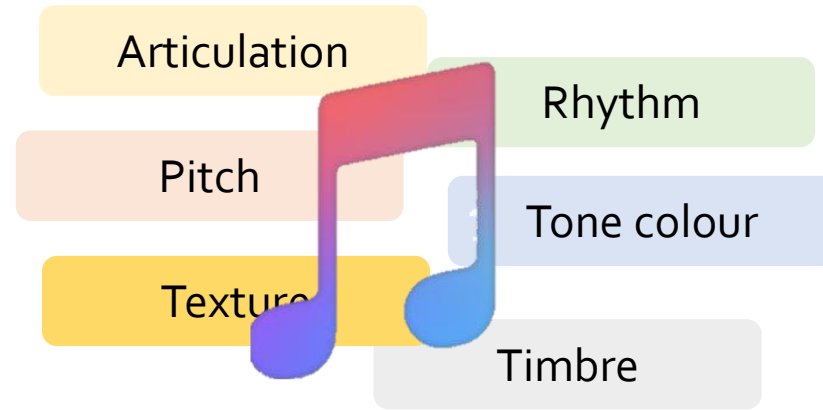
Personalized sleep music:

Automatically bring therapeutic properties into arbitrary user-preferred music



# Challenges

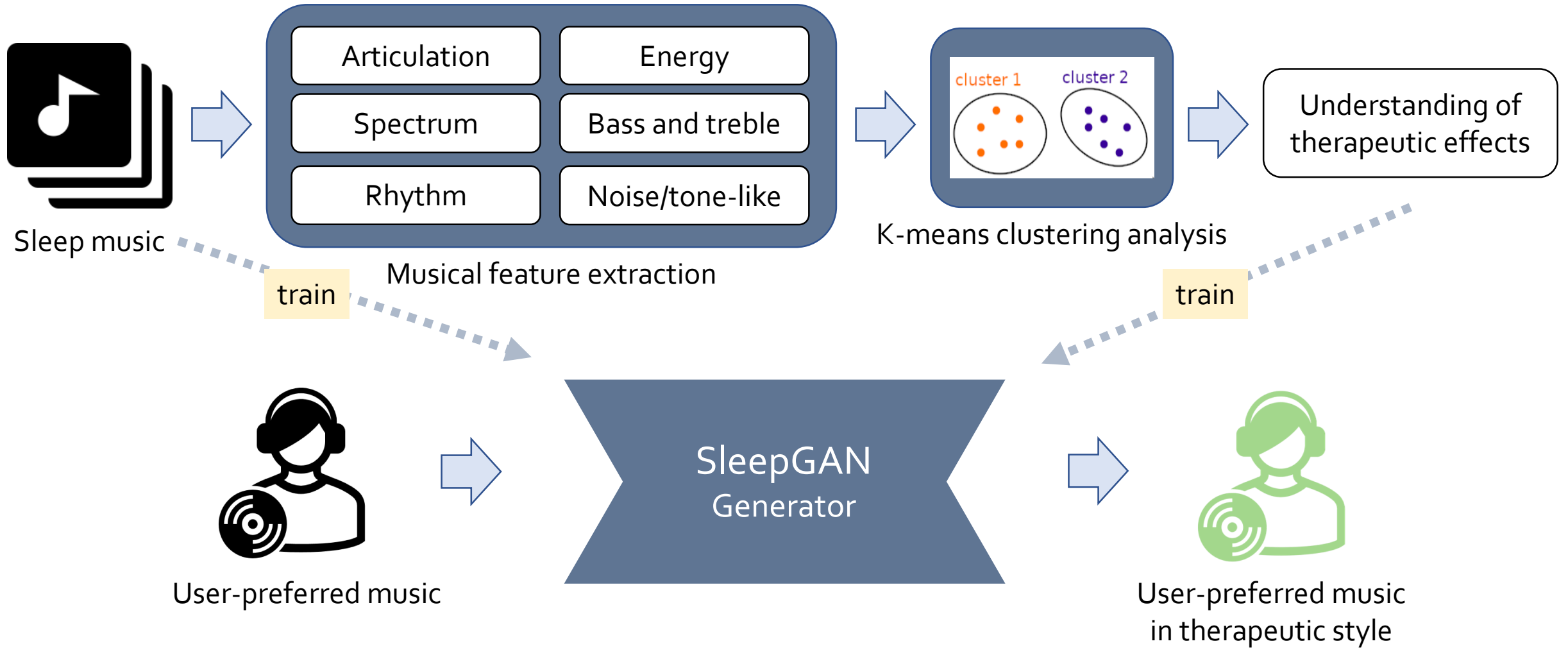
Why so difficult to generate personalized sleep music?



RQ1: What musical features contribute the therapeutic effects of sleep music?

RQ2: How to bring such therapeutic sleep features into user-selected music?

# Overview of Our Approach

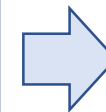
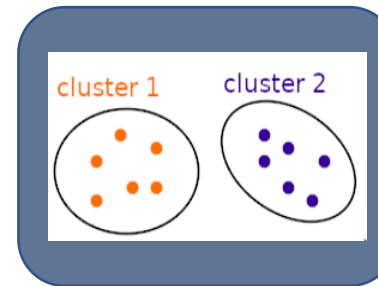
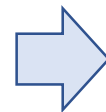
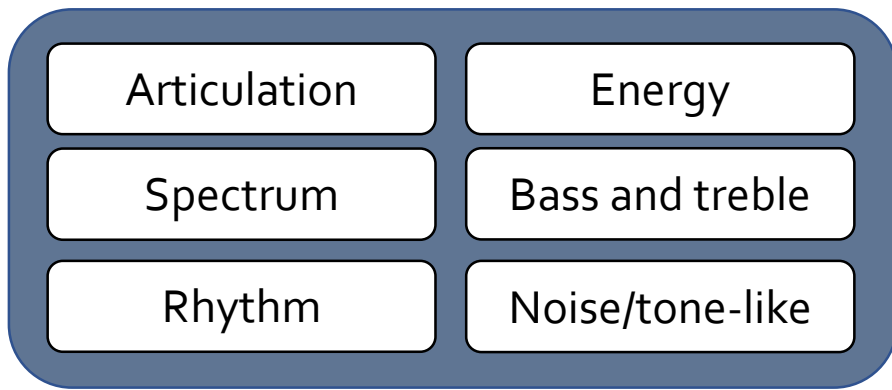
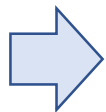




# Musical Feature Analysis



Sleep music



Understanding of therapeutic effects



Other music styles (e.g. jazz, pop, metal)

$$V_{musical} = [f_1, f_2, \dots, f_{33}, f_{34}]$$



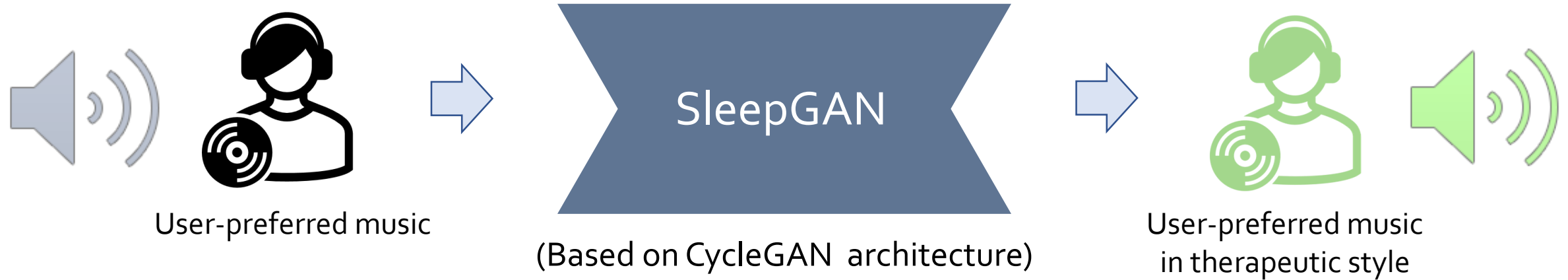
Adjusted Rand Score (ARS):  
Clustering accuracy

	ARS
All 34 musical features	0.115
Only articulation and energy features	-0.063
Only MFCC features	0.096
Only rhythm features	0.112
Only spectral rolloff features	0.761
Only spectral flatness features	0.037

Bass, treble, overall pitch profile

$$w = [w_1, w_2, \dots, w_{33}, w_{34}]$$

# Therapeutic Style Transfer



$$L_{GAN} + L_{cycle} + L_{id} + L_{musical}$$

$$L_{musical} = f(V_{musical} \times \vec{w})$$

$$V_{musical} = [f_1, f_2, \dots \dots f_{33}, f_{34}]$$



# Therapeutic Style Transfer: Evaluation

Objective evaluation target: model-converted music becomes **closer** to sleep music



Cosine distance on VGGish feature vectors



Model-converted music

Sleep music

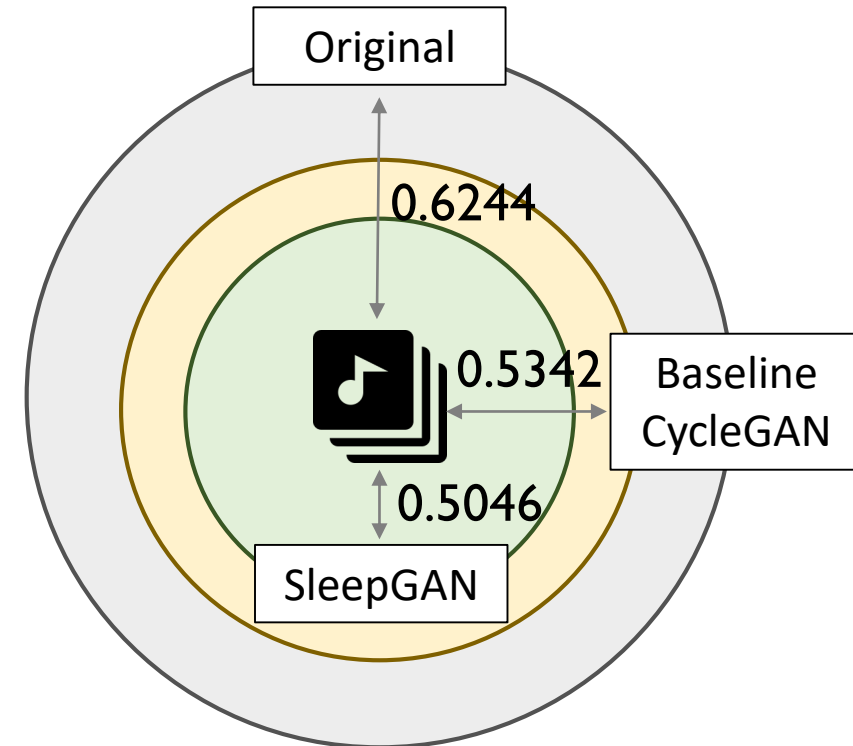
Original music



Our SleepGAN

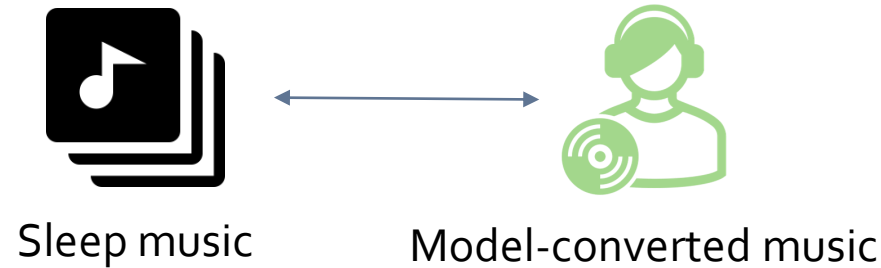
Baseline CycleGAN

Therapeutic version

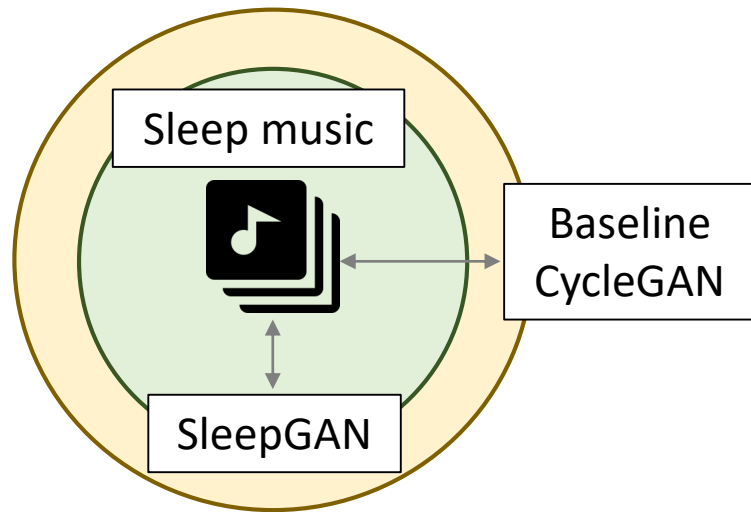


# Therapeutic Style Transfer: Evaluation

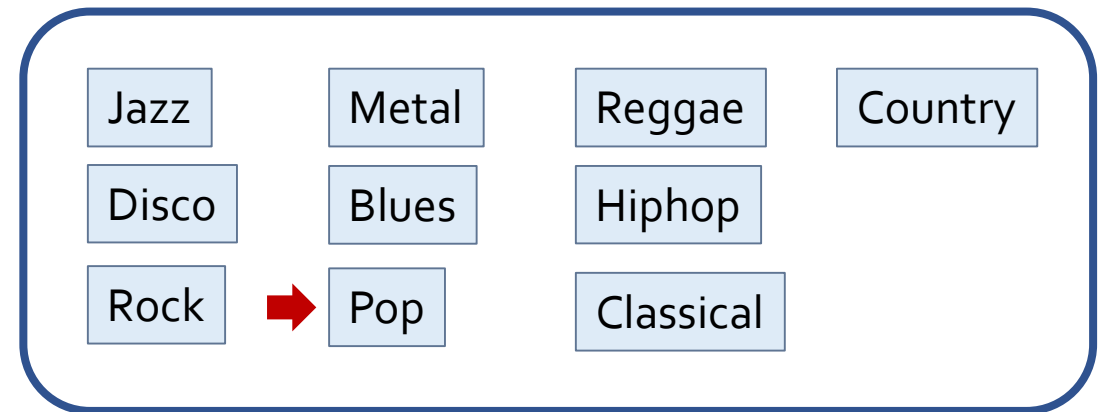
Subjective evaluation with 11 participants (2 female, 9 male)



Perceived style similarity



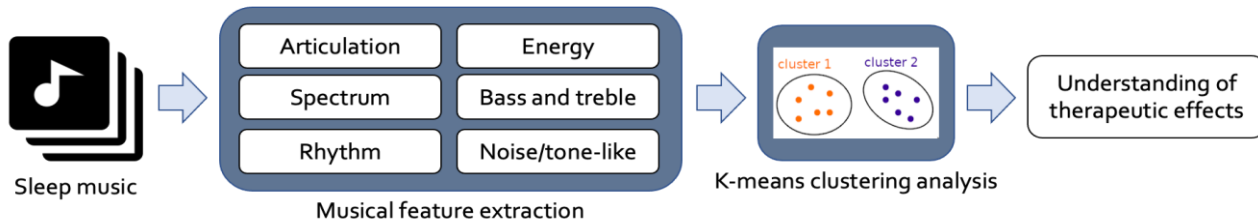
Music genres



More studies to explore the musicality and clinical effects

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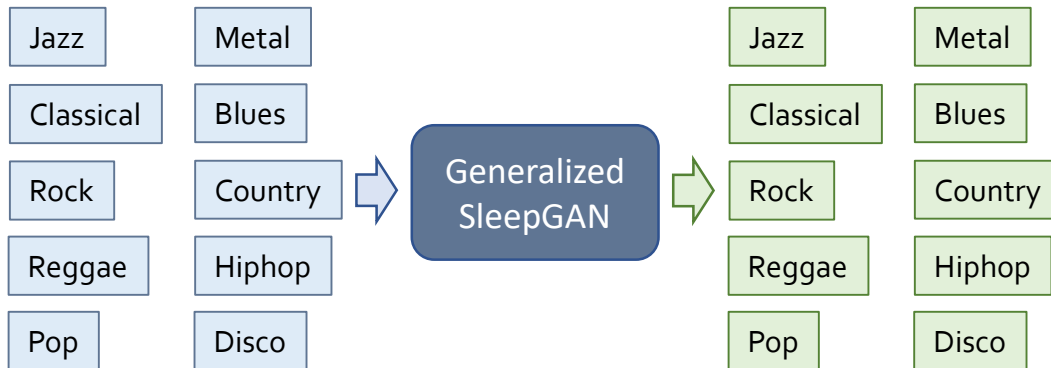
## Take-aways:



**T<sub>1</sub>:** Bass, treble, and overall pitch profile of music

**T<sub>2</sub>:** Music style transfer for creating personal therapy music

## Future work:





# Thank You

## SleepGAN: Towards Personalized Sleep Therapy Music



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