

# Video Game Music Generation with CUDA-trained Markov Chains and Multithreading Optimizations

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15-418 Final Project

## Abstract

Most video games use static soundtracks, which although can be impressive in their own right, can quickly become repetitive to the gamer. In this project, we build a complete system that can process, learn, and generate new video game music. Our system takes in MIDI files, which are processed and trained as a Markov Chain. We then use our Markov Chain to generate a multi-voice music excerpt. We optimize our algorithms by using CUDA techniques in our training algorithm, and multi-threading in the music generation. Speedup analysis shows that our system performs significantly better than sequential algorithms. In fact, because our system is fast, we explore the possibility of integrating our music generation method into video games, so that games can have real-time generated music for a more interactive and unique gaming experience.

## Background

- Music can be treated as a set of rules - ideal for an algorithmic implementation
- Video game music is characterized by the overlapping of multiple melodic lines and a repetitive and structured format
  - Old 8-bit music constrained composers
  - Newer video game music is often symphonic
- Many melodic lines offer ample avenues for parallelization - parallel synthesis of different parts
- Rigid structure makes music easier to predict and generate with a Markov Model

1 <sup>st</sup> - Order Matrix			
Note	A	D	G
A	0.1	0.6	0.3
D	0.25	0.05	0.7
G	0.7	0.3	0

2 <sup>nd</sup> - Order Matrix			
Note	A	C	G
AA	0.18	0.6	0.22
AD	0.5	0.5	0
AG	0.15	0.75	0.1
DD	0	0	1
DA	0.25	0	0.75
DG	0.9	0.1	0
GG	0.4	0.4	0.2
GA	0.5	0.25	0.25
GD	1	0	0

- Matrices built from training data which fit a similar genre
- During generation, probabilities from transition matrix are used to generate next note based on current notes
- Generated music uses structure and patterns from training music
- Training requires looking at millions of notes and updating a very large matrix
- Generation has ~10 music lines generated at once, where each line requires complex computation and synchronization

## Process

### Preprocessing

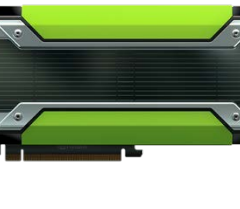
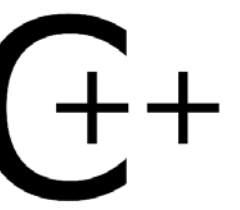
- Convert MIDI to text
- Transpose to C Major
- Specific encoding



music21

### Markov Model Training

- Counts note transitions
- Parallelized over multiple GPUs and threads
- Split between matrix type (major/minor, soprano/bass/chord) on GPUs
- Split notes into sections for many threads to read



### Music Generation

- Creates music based on client request and global settings
- Uses trained Markov Chain to make musical choices
- Multithreading to generate many voices at once
- Synchronization of voices via the conductor



websockets

multi-processing

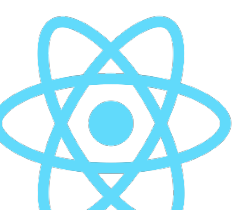
$\begin{bmatrix} 0.5 & \dots & 0.4 \\ \vdots & \ddots & 0.5 \\ 0.1 & \dots & 0.1 \end{bmatrix}$



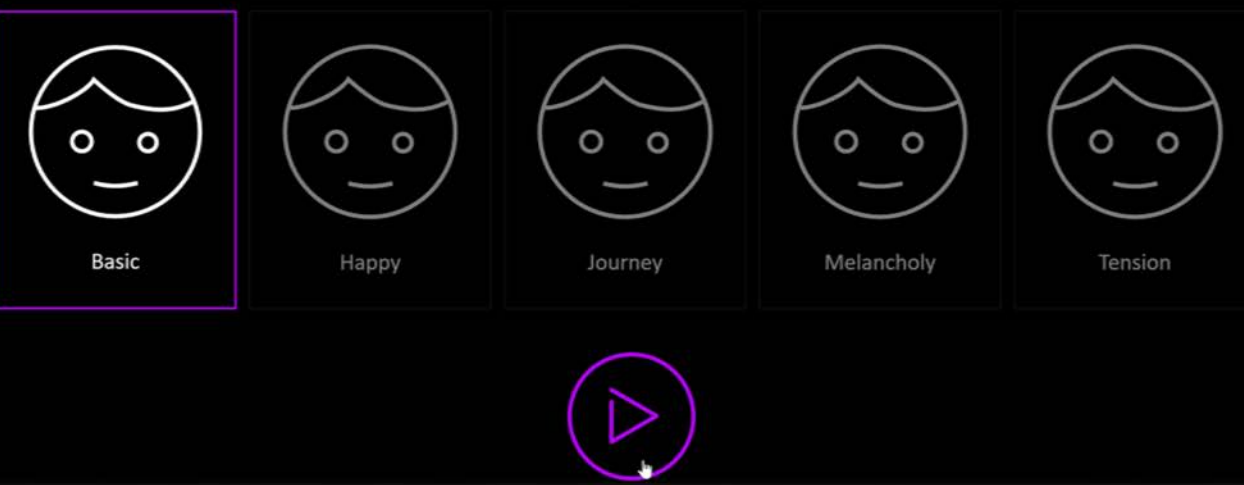
### Client

- Requests music
- Plays notes from server

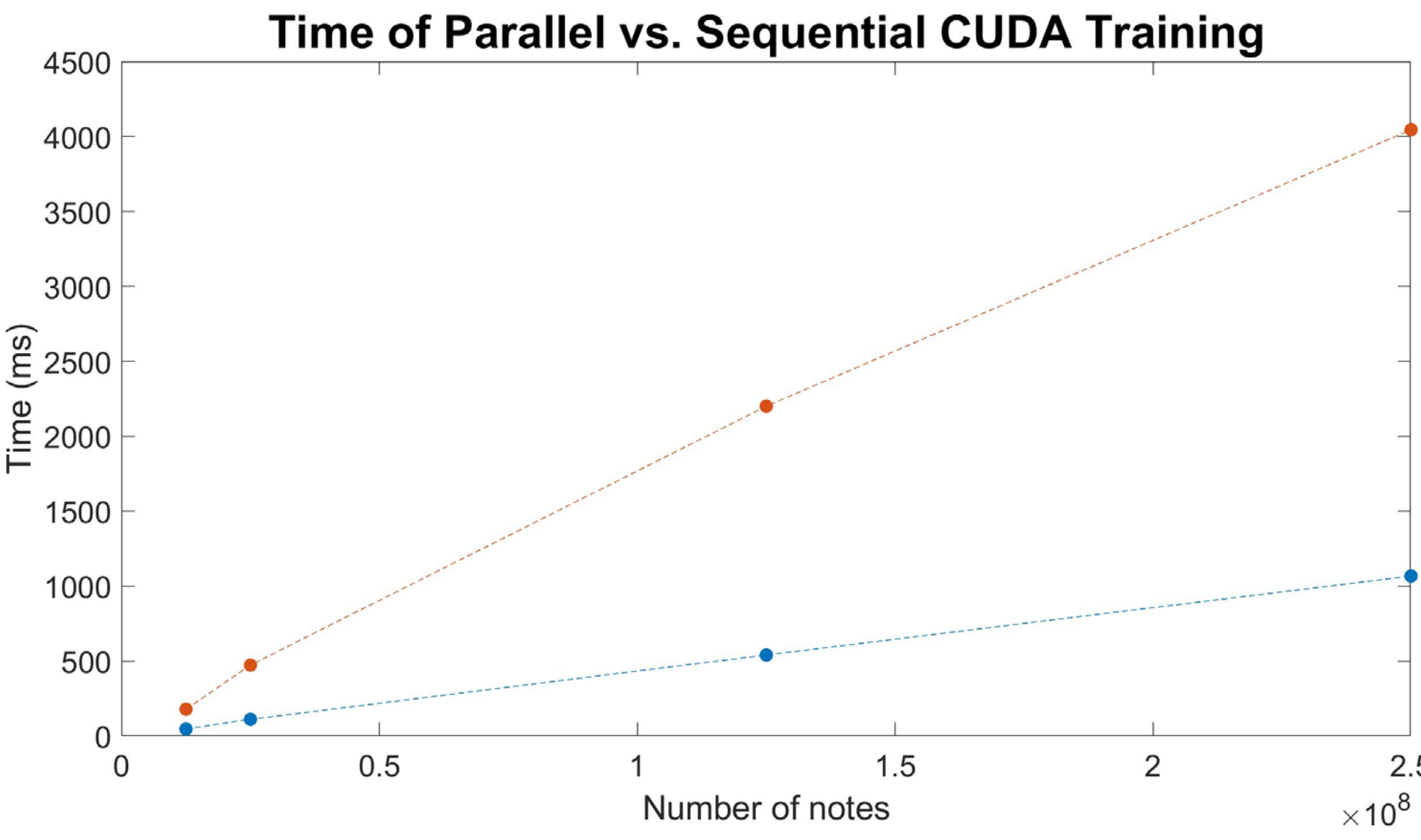
Tone.js



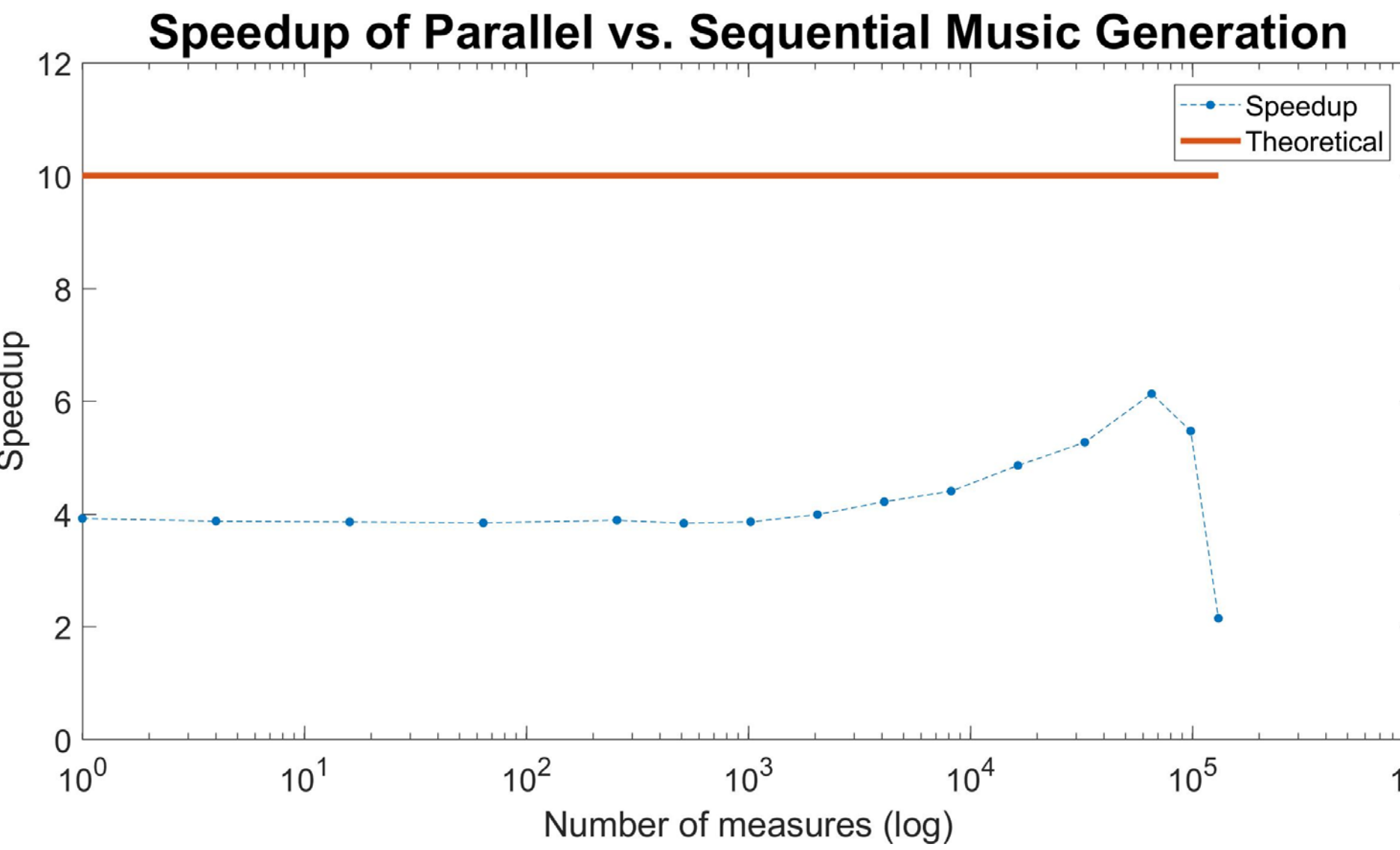
### Parallel Video Game Music



## Optimizations and Results



- Isolated parallel section of algorithm to produce speedup data counting transitions
- Speedup stays steady at 4
- Linear runtime increase in parallel algorithm suggests full thread usage
- CPU-GPU transfer times dominate performance
- Optimizations for async memory copy operations, multiple GPU usage, matrix block divisions



- Speedup increases until 50,000 measures
- Shape of speedup graph indicates balance between output communication and computation
- Matrix loading dominates for low # measures
- Optimizations for CUDA, multi-threading, and multi-processing, as well as matrix loading

### Execution Breakdown

	Parallel Pinned	Parallel Unpinned	Sequential
Allocating Host Memory	10.31047	1.398417	0.000051
Allocating Device Memory	9.545223	3.524767	0
Parsing File and Counting	3.361578	3.370691	8.501543
Copying to Host	0.013167	7.993404	0
Free Memory	4.644862	0.049685	0.045188
Total	27.8753	16.336964	8.546782

## Challenges Faced

- MIDI files are hard to parse and missing important information – key signature, rhythm, etc.
- Required pre-processing and re-formatting of input MIDI files into a text file
- Pre-processing and server algorithms use different languages and parallelism techniques
- Balance of size in Markov Model - larger matrices mean more detailed probabilities and likely better sounding music, but take up a lot of storage
- Client-Server communication of generation settings and live music generation results
- Multithreading vs. GPUs balance in Markov training (much more material to parallelize) and generation (fewer parts and heavier computation)
- Gave user the ability to customize the music they wanted to generate, including live-mixing

## Conclusion

- Created a real-time video game music generator!
- With parallelism, we see speedup in both model training and music generation
- Depending on the structure of the issue, multiple forms of parallelism (GPUs vs. multithreading) are better for different problems
- Even with communication costs, especially between voices, parallelism is still able to achieve both speedup and maintain musicality

## Future Work

- Explore feasibility of using parallel computation for video game music in industry
- Integrating generated game music with gameplay
- Further and more complicated setting manipulation on the client side
- Expansion into other genres/types of music

### Acknowledgments

- Professor Mowry and Professor Railing
- 15-418 TA Staff
- Carnegie Mellon University

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