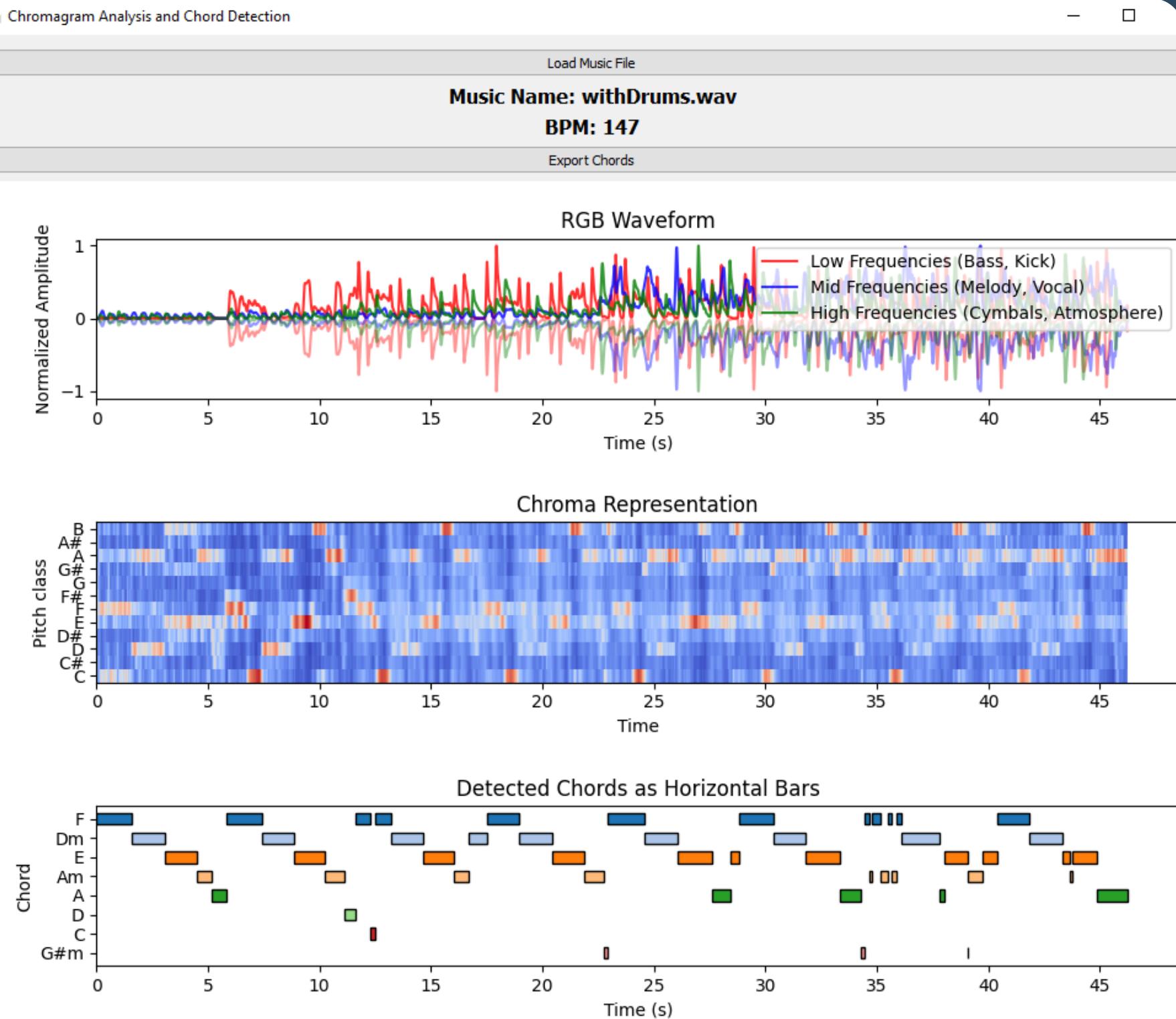


# **CHORD RECOGNITION USING CHROMAGRAM ANALYSIS AND SIGNAL PROCESSING**

**GROUP-9**

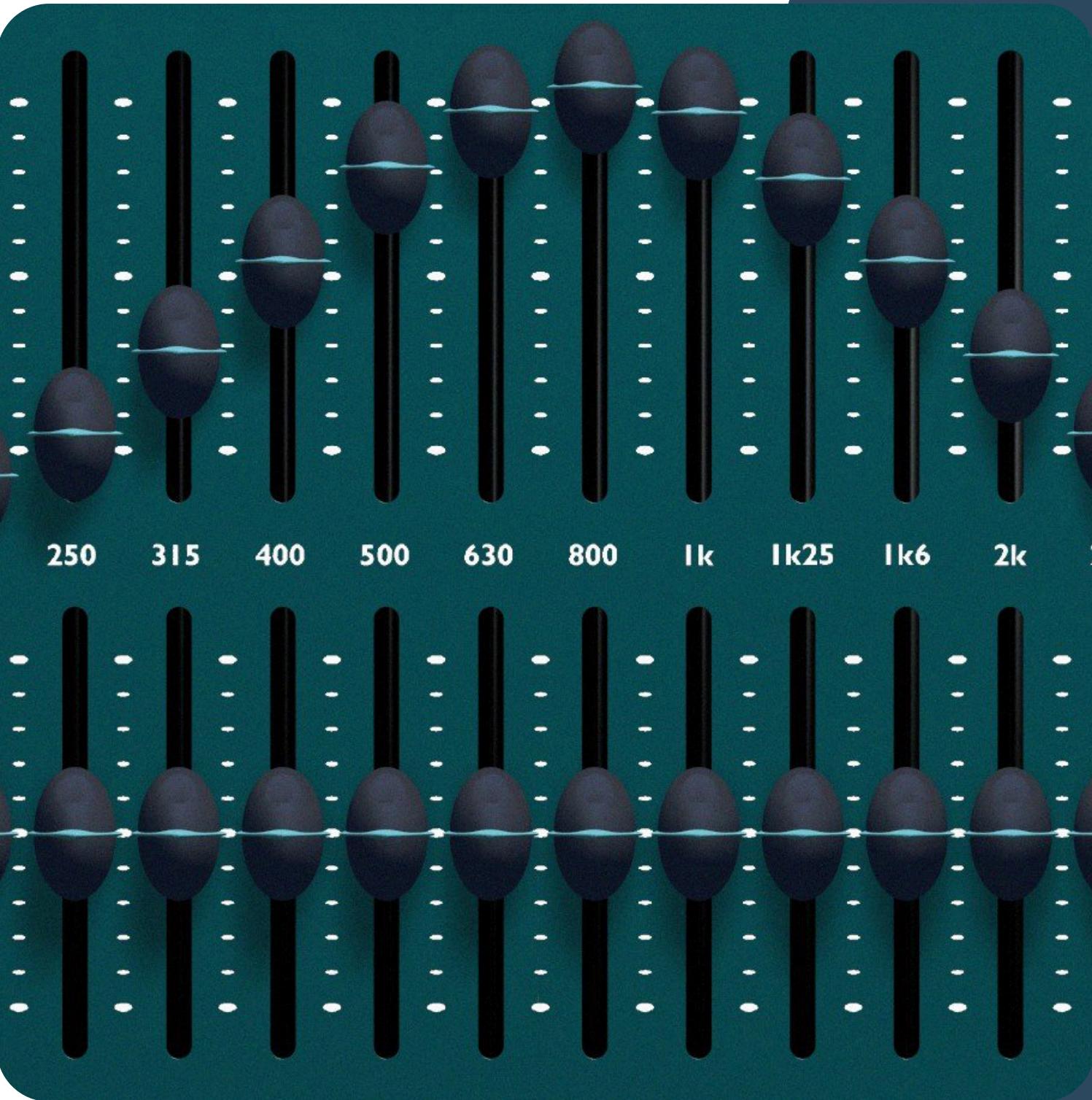
# Introduction

- **Definition of chord**
  - A group of (typically three or more) notes sounded together, as a basis of harmony
- **Importance of Music Analysis**
  - Applications: Automatic transcription, karaoke systems, music recommendation
- **Focus on Chord Recognition**
  - Understanding musical structure
- **Motivation**
  - Growing demand for tools like Chordify
  - Personal interest in the music industry



# Signal Processing & CQT

- Preprocessing Audio:
  - Resampling to 44.1 kHz, converting to mono
- Constant-Q Transform (CQT):
  - Generates chroma features
  - Advantages over STFT for musical signals
- Chromagram Enhancement:
  - Normalization and smoothing to detect chords accurately



# Methodology Overview

## 1st Main Step

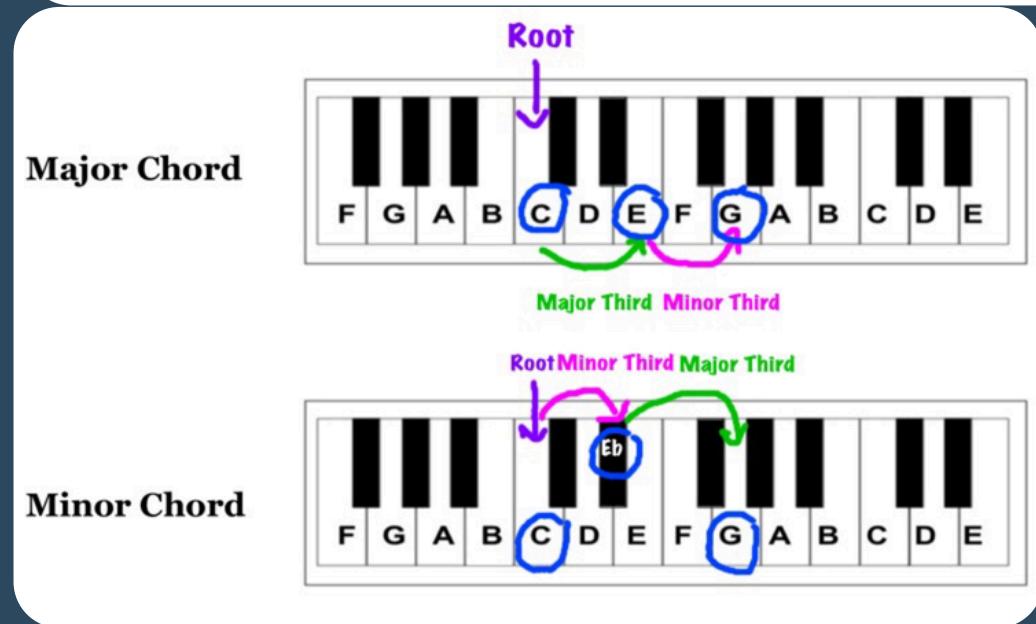
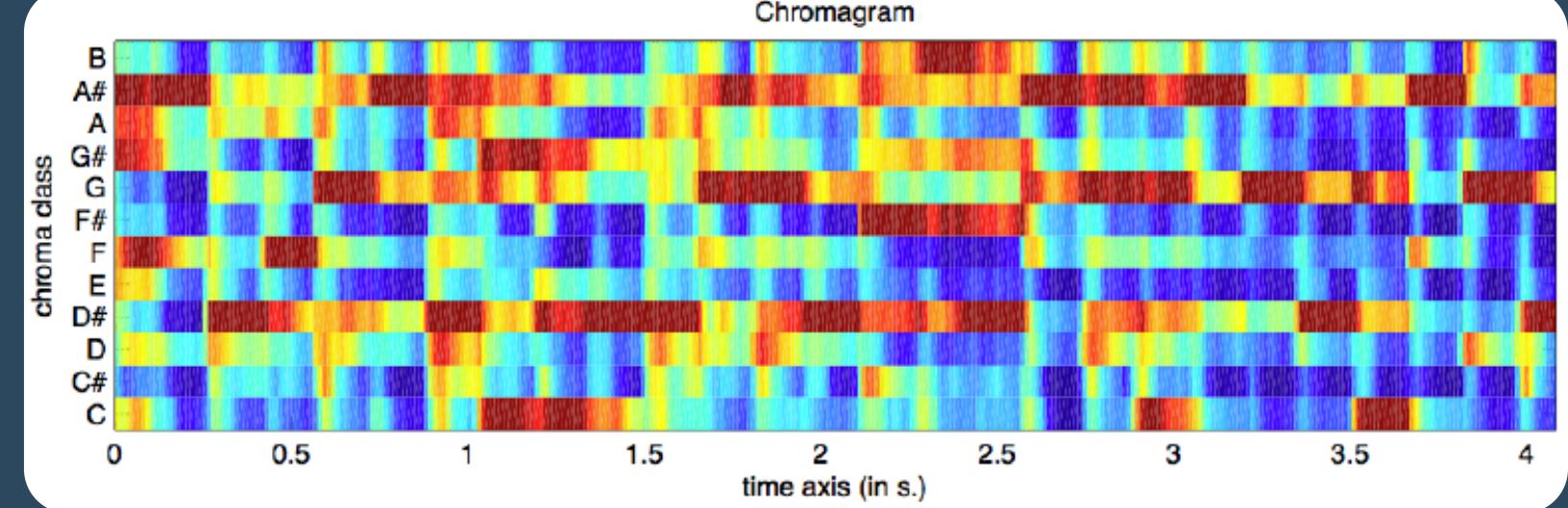
Signal Processing &  
Chromagram Generation

## 2nd Main Step

Chord Detection via  
Templates

## Tools & Libraries

Python, Librosa,  
Matplotlib, PyQt5



# Chord Detection

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## 01. Chord Templates

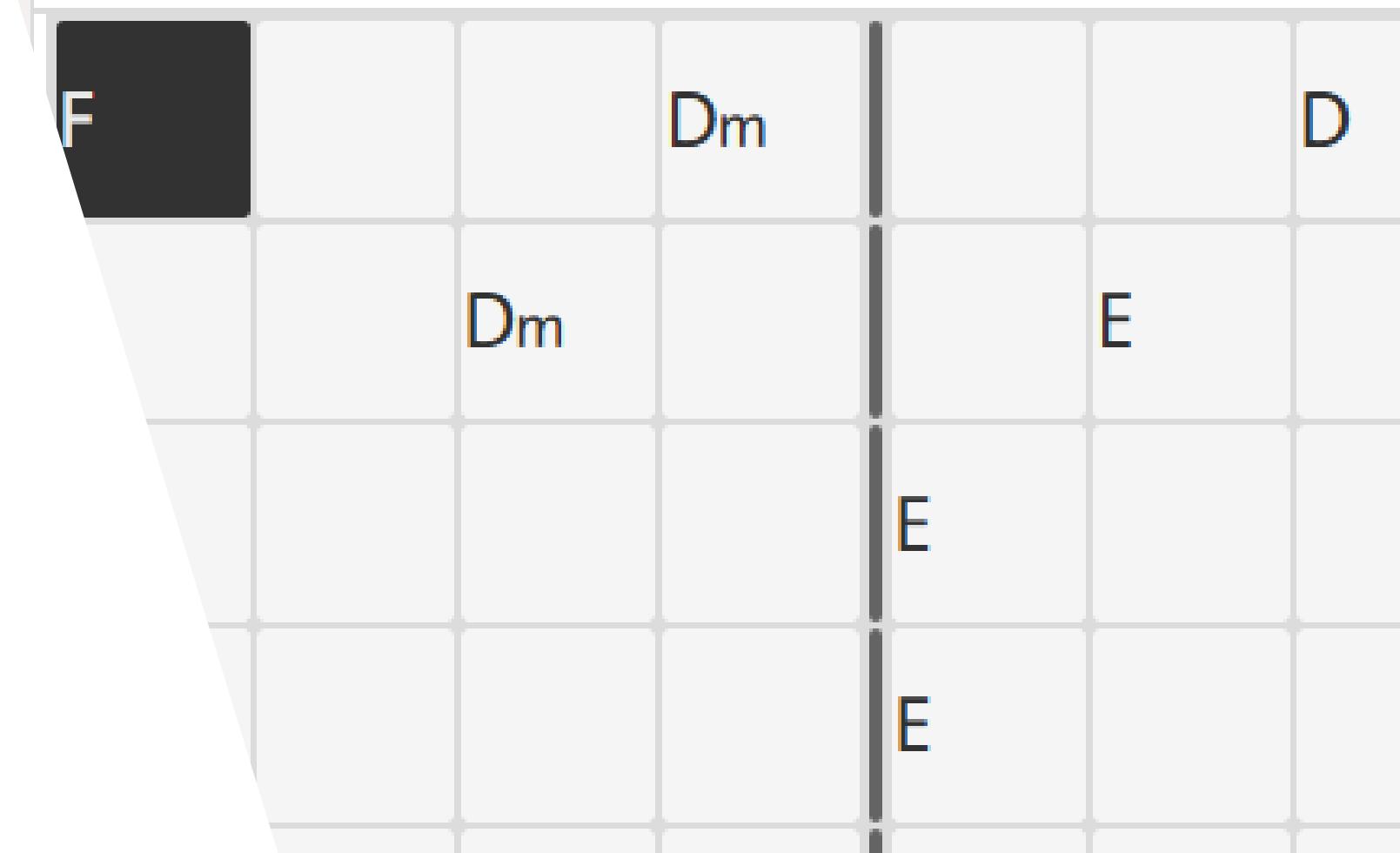
- 12 major and minor chords as 12-element vectors

## 02. Detection Process

- Compare chromagram frames with templates using dot product
- Assign chord with highest similarity

## 03. Smoothing Algorithm

- Applies a window to stabilize chord sequence



Dm

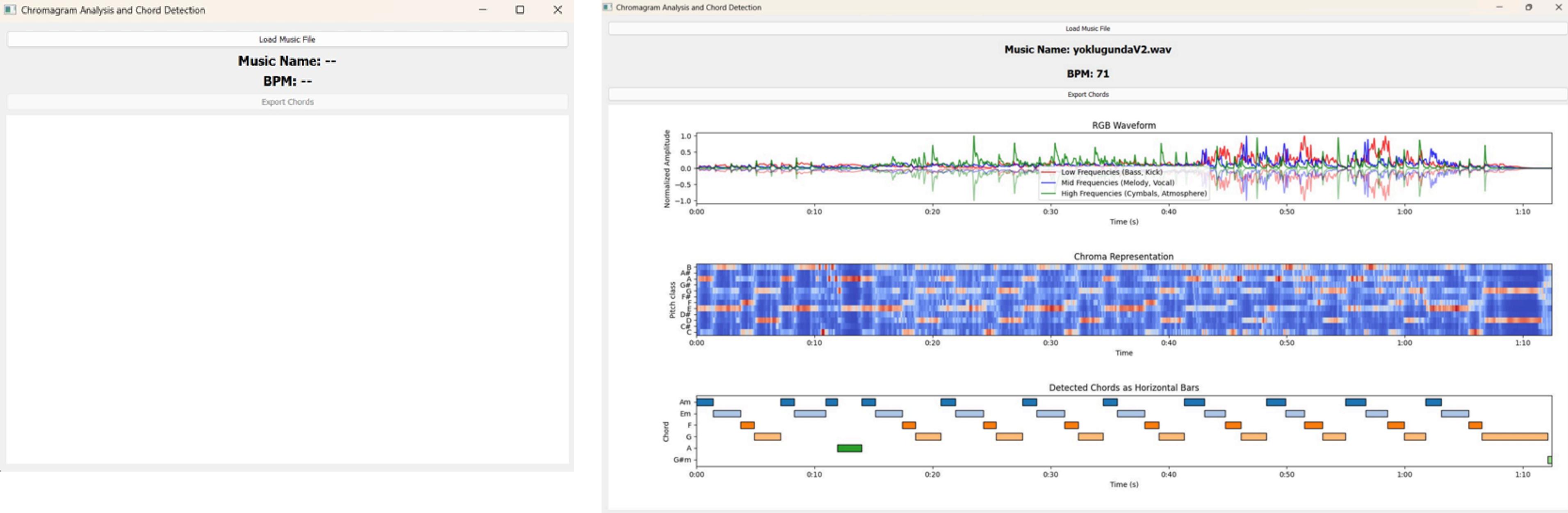
# Implementation & Visualization

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BPM: 71		
Start Time	End Time	Chord
0	1.428027	Am
1.428027	3.738413	Em
3.738413	4.89941	F
4.89941	7.128526	G
7.128526	8.266304	Am
8.266304	10.94821	Em
10.94821	11.94667	Am
11.94667	14.00163	A
14.00163	15.13941	Am
15.13941	17.43819	Em
17.43819	18.54113	F
18.54113	20.67737	G
20.67737	21.94286	Am
21.94286	24.28807	Em
24.28807	25.39102	F
25.39102	27.62014	G
27.62014	28.81596	Am
28.81596	31.1844	Em
31.1844	32.3454	F
32.3454	34.42358	G
34.42358	35.61941	Am
35.61941	37.94141	Em
37.94141	39.14884	F
39.14884	41.31991	G
41.31991	43.03819	Am

◦ Figure 3: Exported chords in CSV

- Implementation Steps
  - Load and preprocess audio files
  - Compute chromagram using Librosa
  - Detect chords and apply smoothing
  - Visualize with Matplotlib
- GUI with PyQt5:
  - User-friendly interface for loading files and exporting results
- Visuals:
  - Figure 1: Project main menu
  - Figure 2: Processed music file
  - Figure 3: Exported chords in CSV



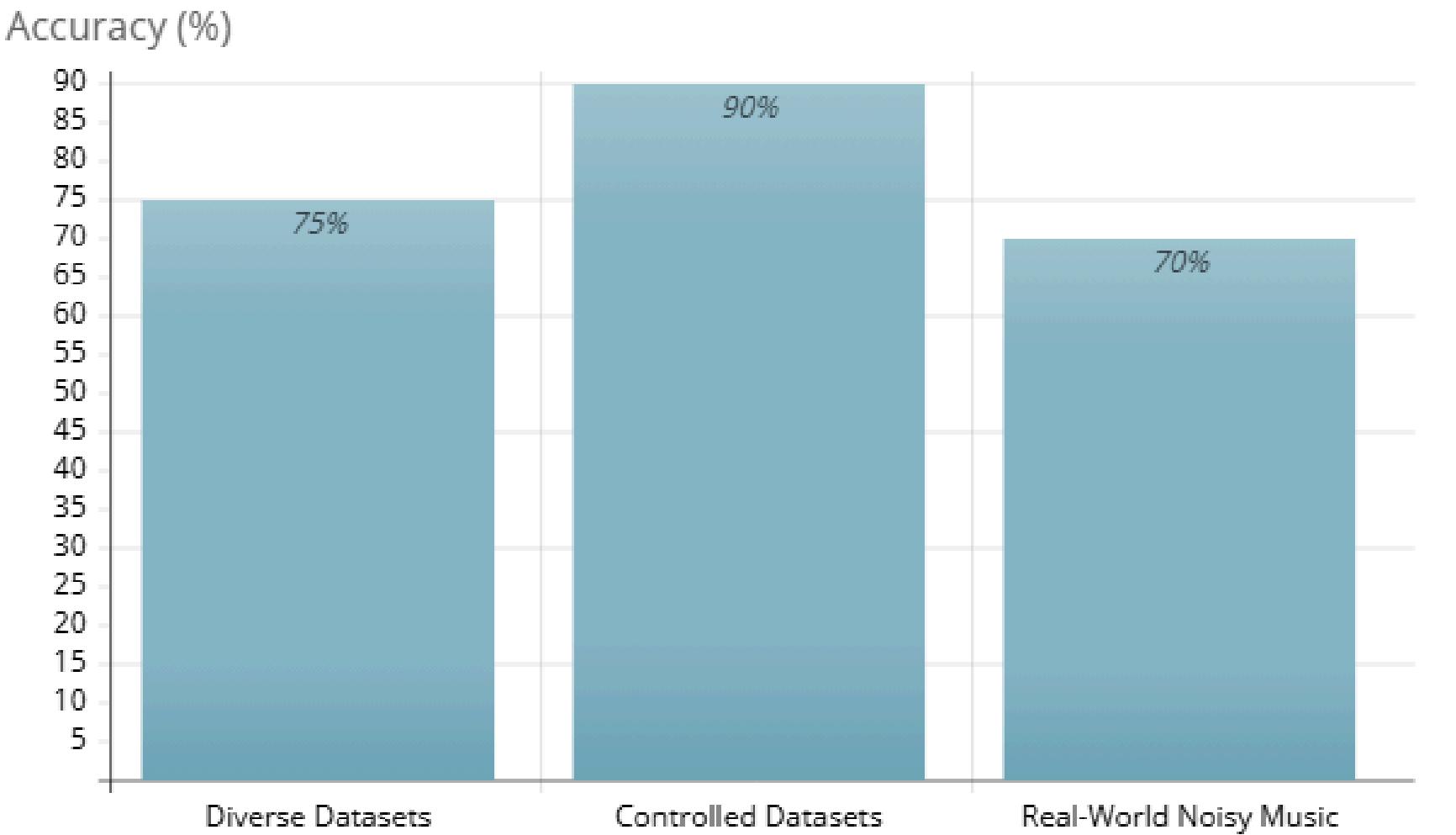
◦ Figure 1: Project main menu

◦ Figure 2: Processed music file

# Results & Discussion

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- Accuracy Rates
  - 65–75% on diverse datasets
  - Up to 85–90% on controlled datasets
  - 50–70% on real-world noisy music
- Comparison:
  - Advanced ML algorithms achieve ~95% accuracy
  - Achieved high accuracy with basic signal processing
- Challenges:
  - Noise and complex music structures





# Traditional Waveform



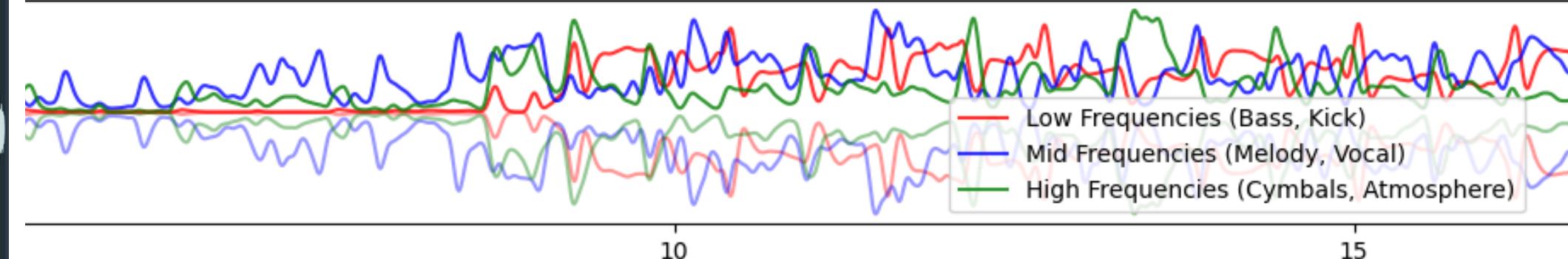
**SERUM(VST)**



**FL Studio (DAW)**



# Our Special Waveform



**Voxengo SPAN**



# References

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- Brian McFee et al., "librosa: Audio and Music Signal Analysis in Python," Proceedings of the 14th Python in Science Conference, 2015.
- librosa Documentation
- matplotlib Documentation
- PyQt5 Documentation
- M. Mueller, Fundamentals of Music Processing Audio, Analysis, Algorithms, Applications. Springer Verlag, 2015.
- Caio Miyashiro, Hidden Markov Models for Chord Recognition, Pydata 2019 Conference

```
        }
    for i, note in enumerate(notes):
        _template = np.zeros(12)
        intv = major_intervals[i]
        major_template[(i + intv) % 12] = 1
        chords[f"{n}"] = major_template

        minor_template = np.zeros(12)
        if intv in minor_intervals:
            minor_template[(i + intv) % 12] = 1
        chords[f"{n}m"] = minor_template
    n_chords

# Detection function
def chord_detection(chroma, smoothing_window):
    chords = get_chord_templates()
    chord_names = list(chords.keys())
    chord_templates = np.array([chords[ch] for ch in chord_names])
    chord_templates = chord_templates / np.linalg.norm(chord_templates, axis=1)

    detected_chords = []
    for frame in chroma.T:
        similarities = chord_templates.dot(frame)
        best_chord_index = np.argmax(similarities)
        detected_chords.append(chord_names[best_chord_index])

    # Smooth chord sequence to remove rapid fluctuations
    if len(detected_chords) > 0:
        smoothed_chords = []
        for i in range(len(detected_chords)):
            start = max(0, i - smoothing_window)
            end = min(len(detected_chords), i + smoothing_window + 1)
            window_chords = detected_chords[start:end]
            # Choose the most common chord in the window
            most_common_chord = Counter(window_chords).most_common(1)[0][0]
```