

# Audio Features: MFCC & Chroma



## MFCC

Mel-Frequency Cepstral Coefficients

Models human auditory perception

**Based on:** Mel-scale (perceptually motivated frequency scale)

- Speech recognition
- Speaker identification
- Audio classification



## Chroma Features

Pitch Content & Harmonic Structure

Represent pitch content and harmonic structure

**Focus on:** 12 pitch classes (chromatic scale)

- Music analysis
- Chord recognition
- Similarity detection



Both features capture complementary aspects of audio signals



Compact representations suitable for machine learning models



# Visual Process Flow

## MFCC Extraction Process

### 1 Audio Signal

Input original audio signal (time domain)

### 2 STFT

Frequency analysis with Short-Time Fourier Transform

### 3 Mel Filter Bank

Apply Mel-scale filters that reflect human auditory characteristics

### 4 DCT

Extract coefficients with Discrete Cosine Transform

### MFCC Coefficient Example (13 coefficients)



## Chroma Extraction Process

### 1 Audio Signal

Input original music signal

### 2 STFT

Analyze frequency spectrum

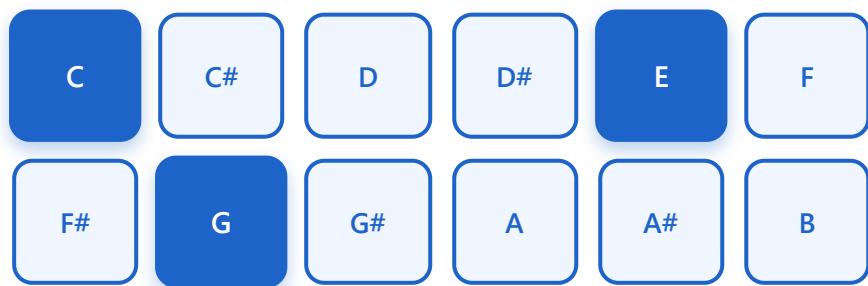
### 3 Pitch Mapping

Map frequencies to 12 pitch classes (C, C#, D, ...)

### 4 Octave Folding

Fold all octaves into one (octave invariance)

### Chroma Vector Example (12 pitch classes)



*Each coefficient represents different frequency characteristics of the audio*

*Highlighted notes = C Major chord (C, E, G)*

*The intensity of each pitch class is represented as a vector value*