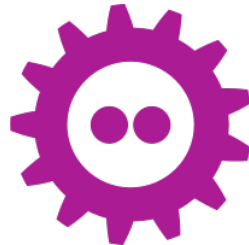


Over-the-air Audio Identification

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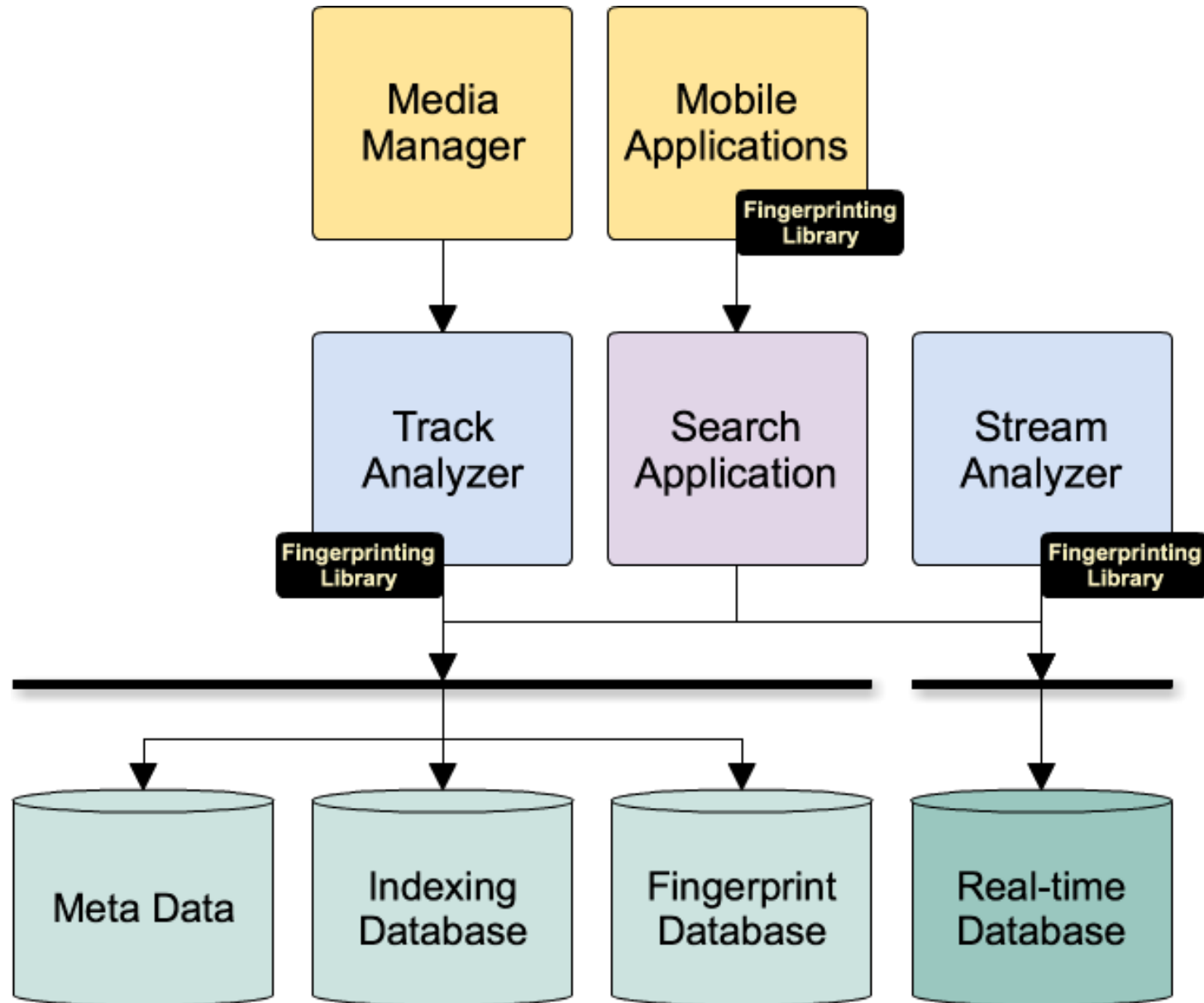
[ardayalciner](https://www.linkedin.com/in/ardayalciner)

OTA Audio Identification

Matching an audio sample with a pre-recorded sound clip

- **Music track** recognition
- **Radio / TV station** detection
- **Licensing**
- **Second screen** applications
 - Previously on *<insert TV Show here>*
 - Track watched movies / TV shows
 - Nearby concerts of playing artist
 - Information on a currently speaking movie / TV show character

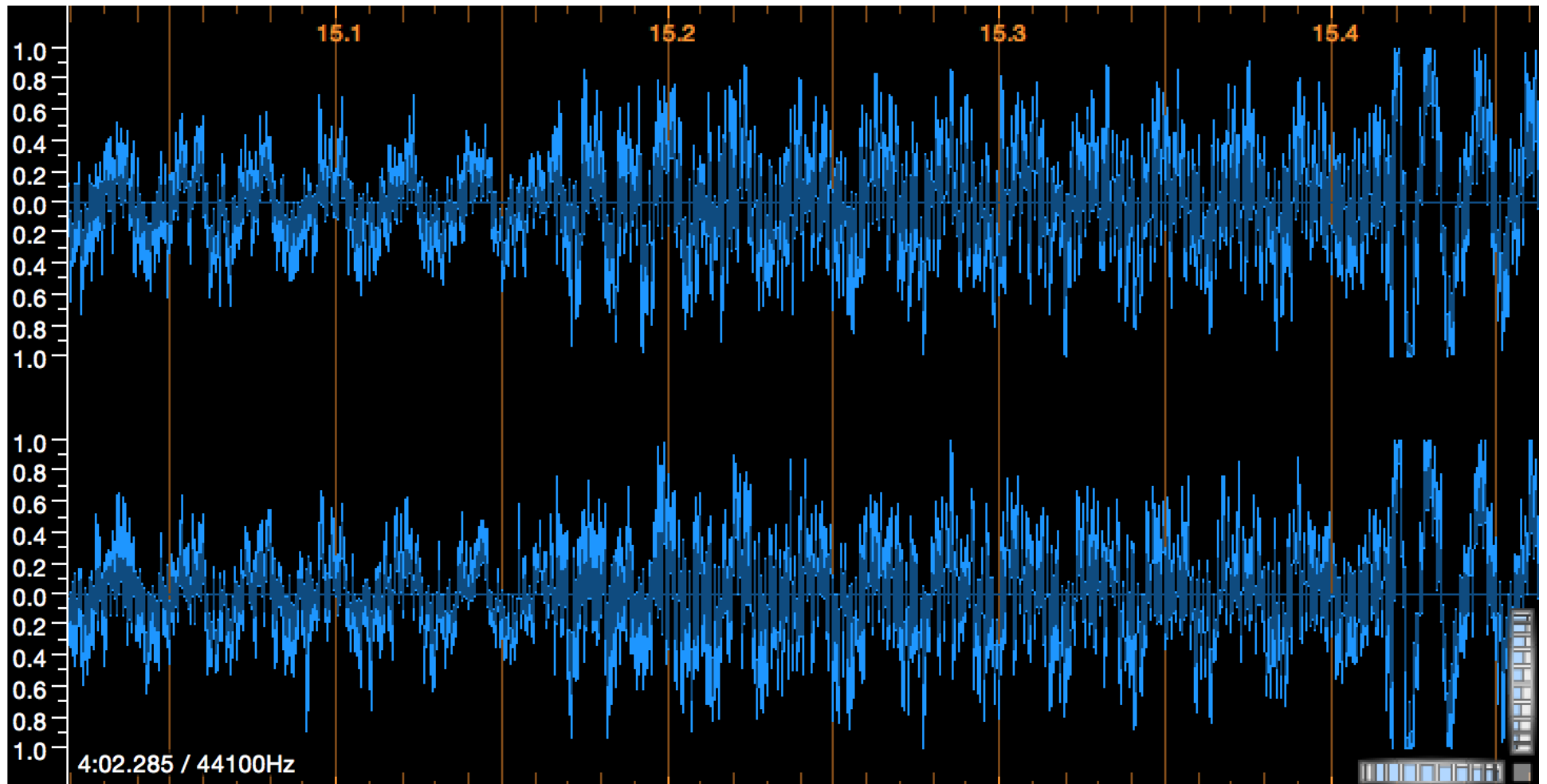
Reference Architecture



Digital Sound Signals

- In nature, sound propagates as **sound waves**.
- We measure **sound pressure** at specific intervals. This interval is called **sample rate**.
- A sample rate of 44.1 kHz means, we measured the sound pressure 44100 times per second.
- These discrete signals represent sound in a digital form.

Digital Sound Signals



Digital Sound Signals

- **Properties:**

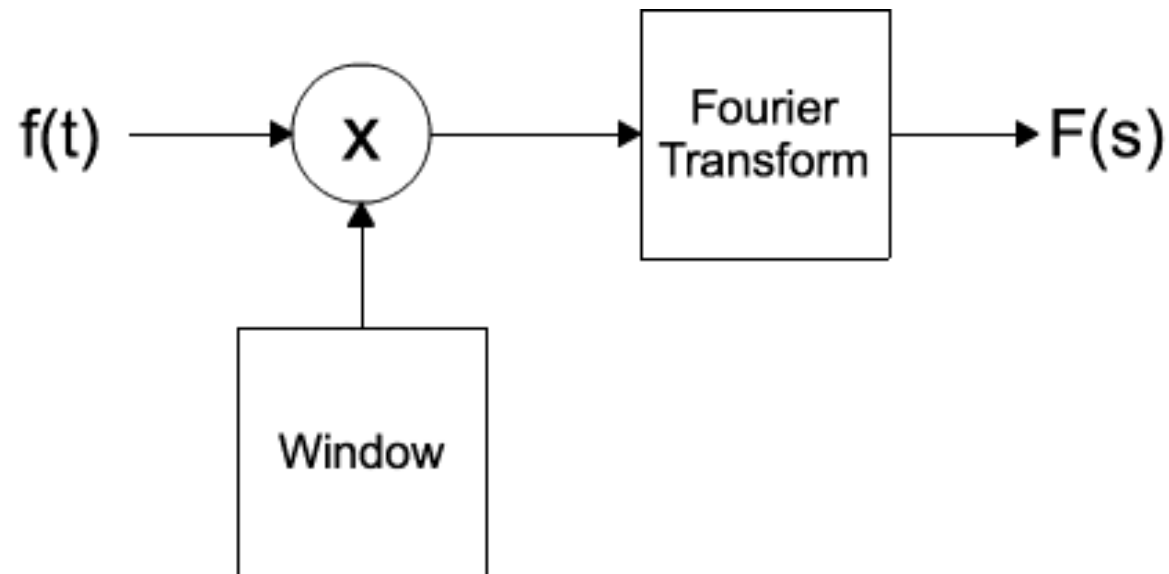
- **Bit depth:** # of bits a sample occupies
- **Channels:** # of simultaneous recordings
(*1: mono, 2: stereo, etc.*)
- **Endianness:** Big-endian vs. Little-endian

- **File Formats:**

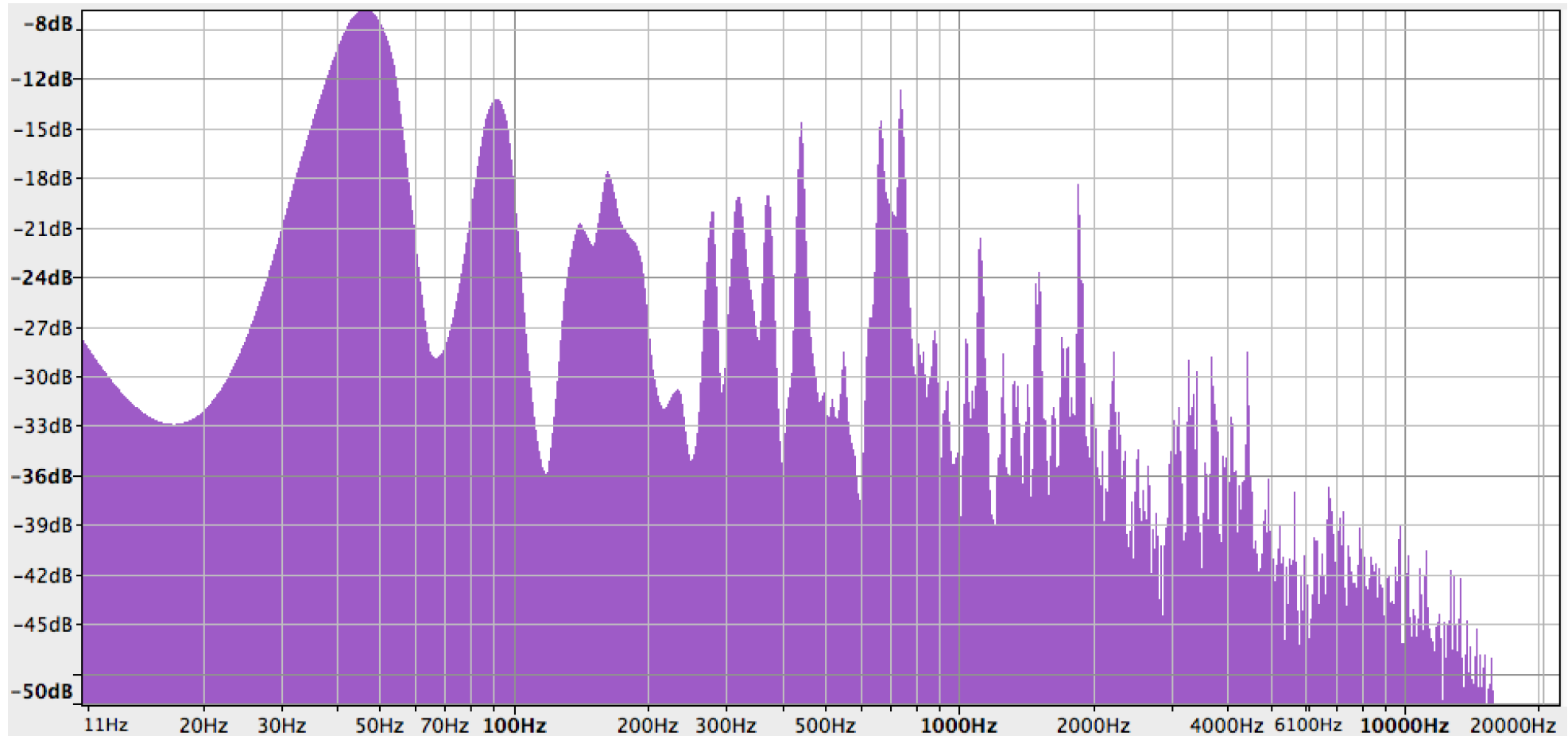
- Uncompressed: **PCM, Wave**
- Compressed:
 - Lossless: **FLAC**
 - Lossy: **MP3, AAC, Ogg**

Frequency Analysis

- Record or play audio signals in the *time domain*:
SPL vs. Time
- Analyze audio signals in the *frequency domain*:
Frequency vs. Amplitude vs. Time

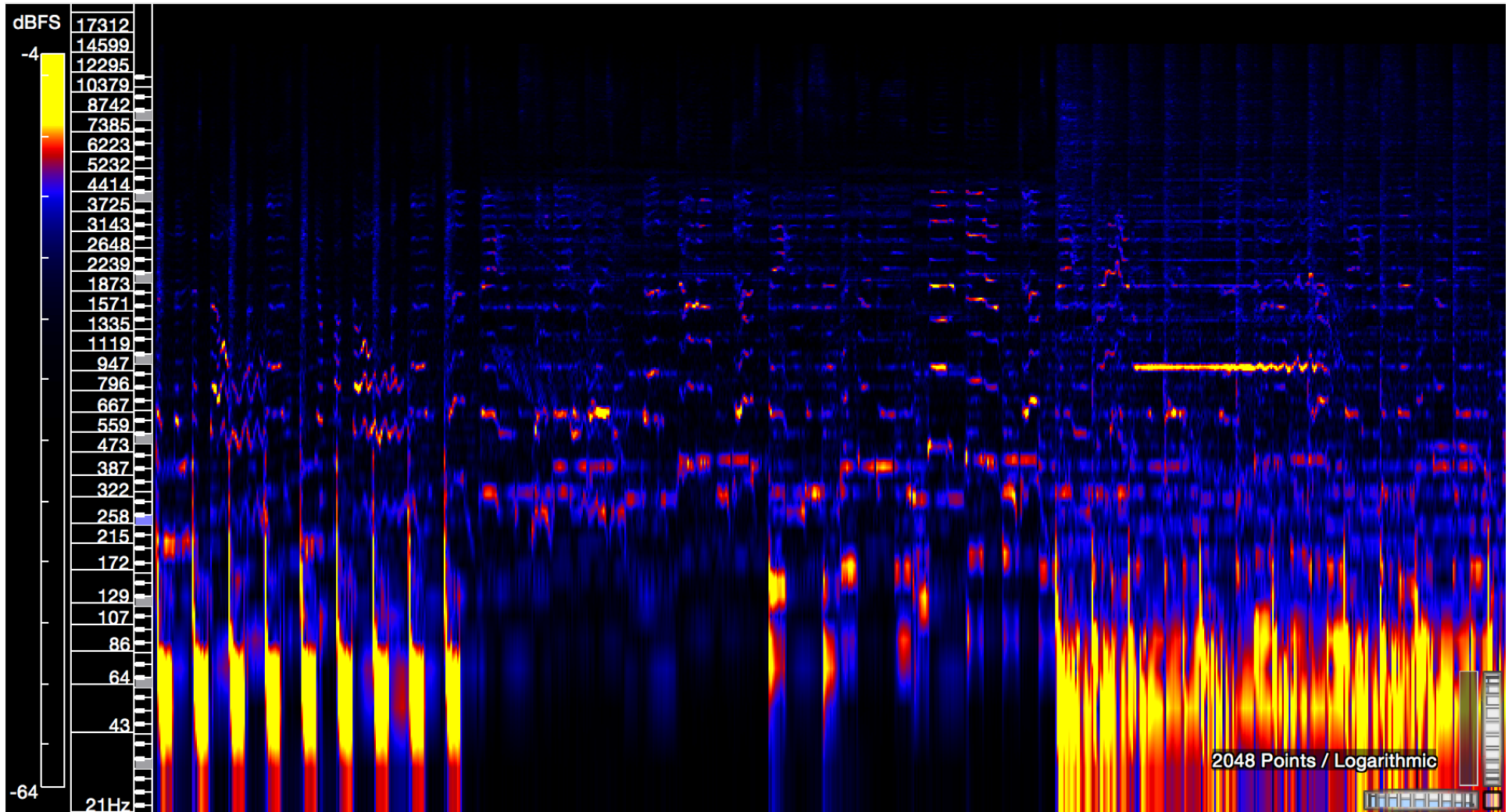


Frequency Analysis: Spectrum



- Covers frequencies up to $0.5 * \text{sample_rate}$ [Hz]
- Divided into bins. Each bin represents the average amplitude for $0.5 * \text{sample_rate} / \text{fft_points}$ wide of frequencies

Frequency Analysis: Spectrogram



- Sensitive either in time dimension or frequency dimension: not both

Fingerprinting

Problem:

We need to **uniquely** summarize a **part of** an audio recording despite various **challenges**

Approach Using:

- Music information retrieval (*MIR*)
- Acoustic fingerprinting

Fingerprinting: MIR

“What can we retrieve?”

More **specific**:

- Musical features (*notes, chords, harmony, rhythm, ...*)
- Speech
- Instruments
- Melody: *Query by Humming*

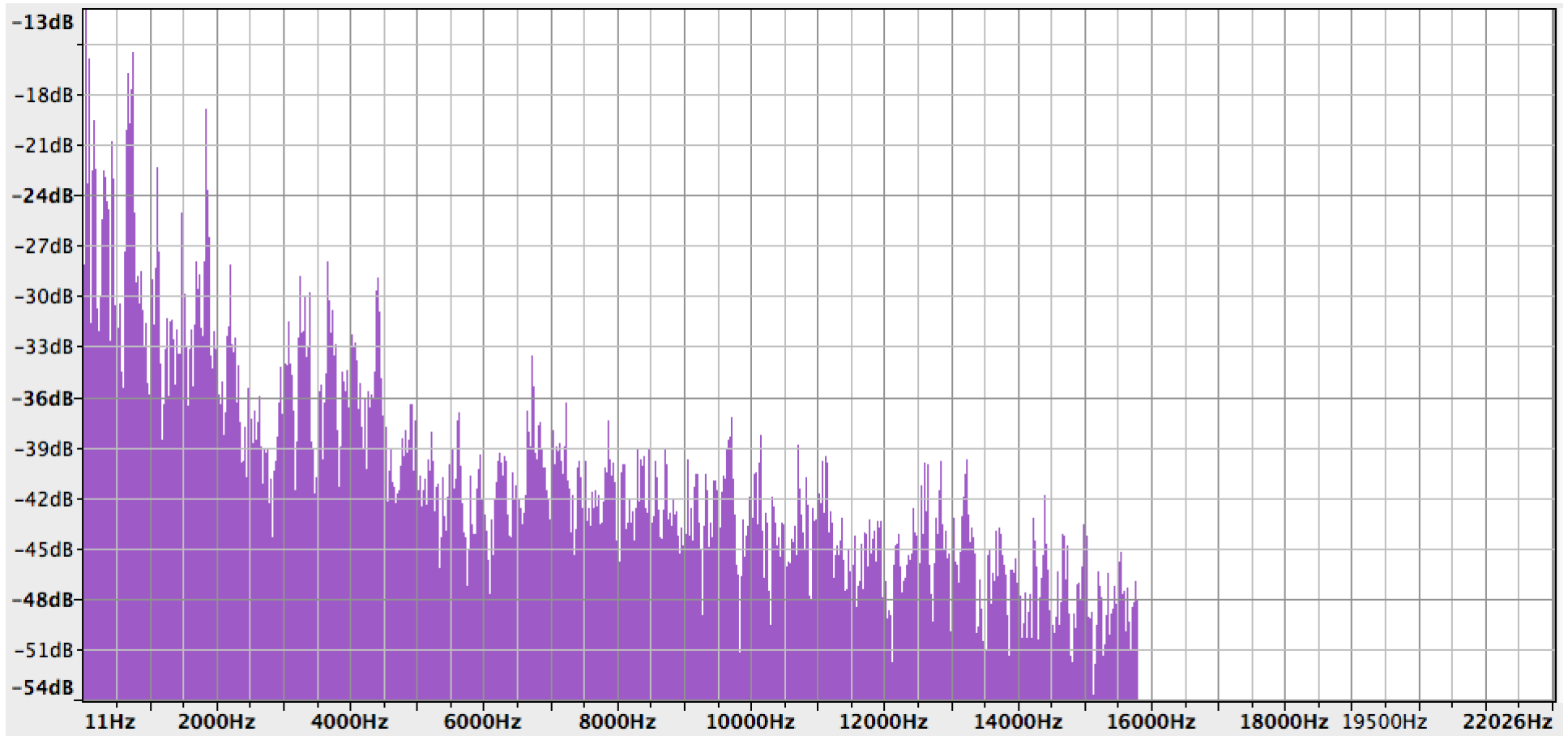
More **abstract**:

- Time-frequency peaks

Fingerprinting: Challenges

- Noise
 - **Duration:** *instantaneous / continuous*
 - **Frequency range:** *small / wide*
 - **Loudness:** *quiet / loud*
- Echo
- Changes in tempo
- Changes in pitch
- Attenuation or boost in certain frequencies
(*e.g., Equalization*)

Fingerprinting: Time-Frequency Peaks



- **Divide** the spectrum into **N** equal areas (e.g., 16 parts)
- For each area, find the **frequency bin** that provides the **peak amplitude**

Fingerprinting: Packing

FFT Points	P = 1024
# of Areas	N = 16
# of Bins / Area	0.5 * P / N = 32
Sample Rate	SR = 11025
Max. Frequency	SR / 2 = 5513

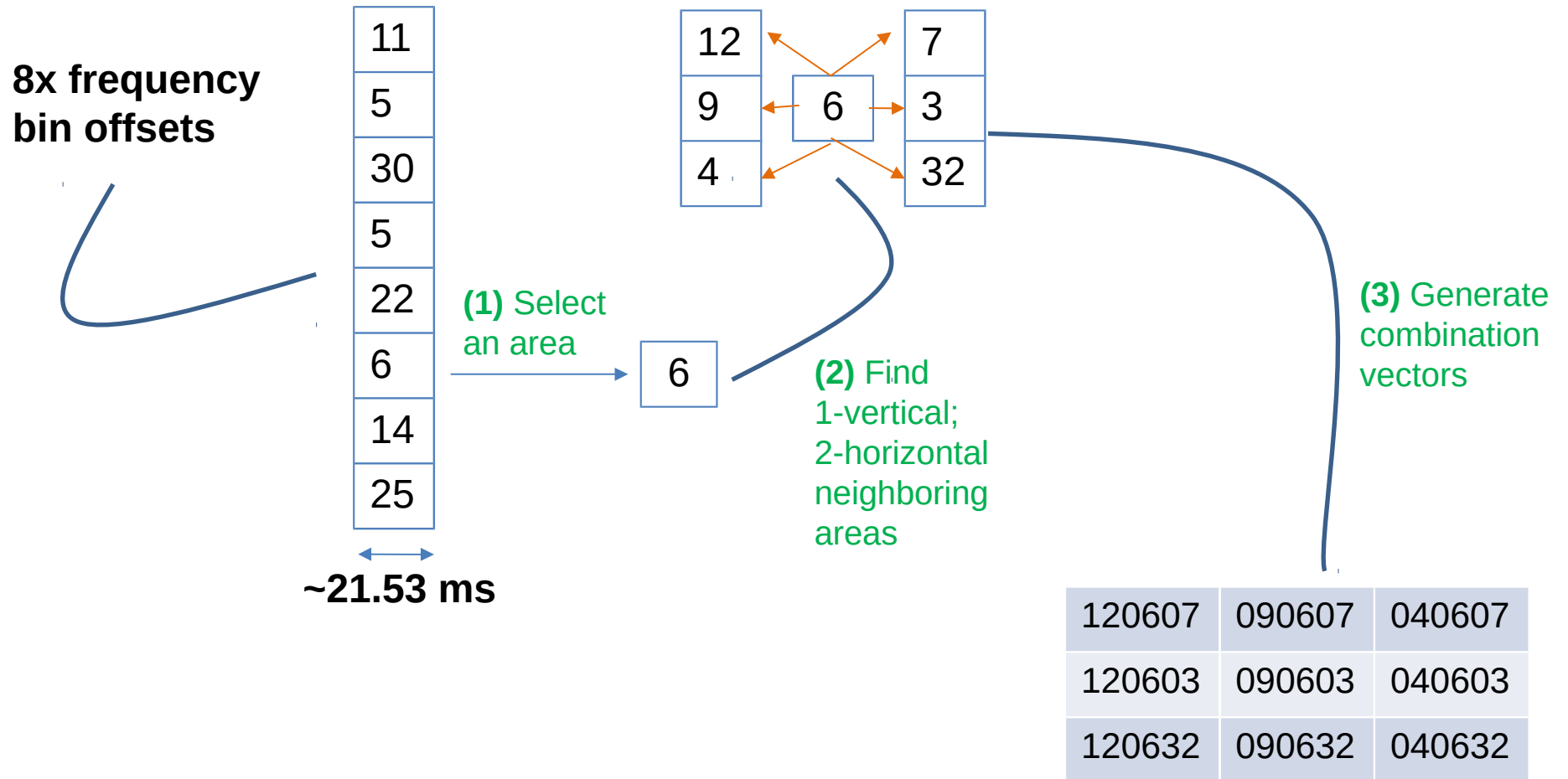
We can represent 5513 using a 16-bits integer.
16 of them occupies **256-bits** (32 bytes).

However, we can represent 32 with 5-bits.
It is possible to store them in **80-bits** (10 bytes).

i	0	1	2	3	4	5	14	15
F	269	495	753	1270	1431	2045	4876	5285
b	25	14	6	22	5	30	5	11

$$fp = \sum_{i=0}^{16-1} (b[i] - i * 32) * 2^{5*i}$$

Fingerprinting: Hashing



Fingerprinting: Key Choices

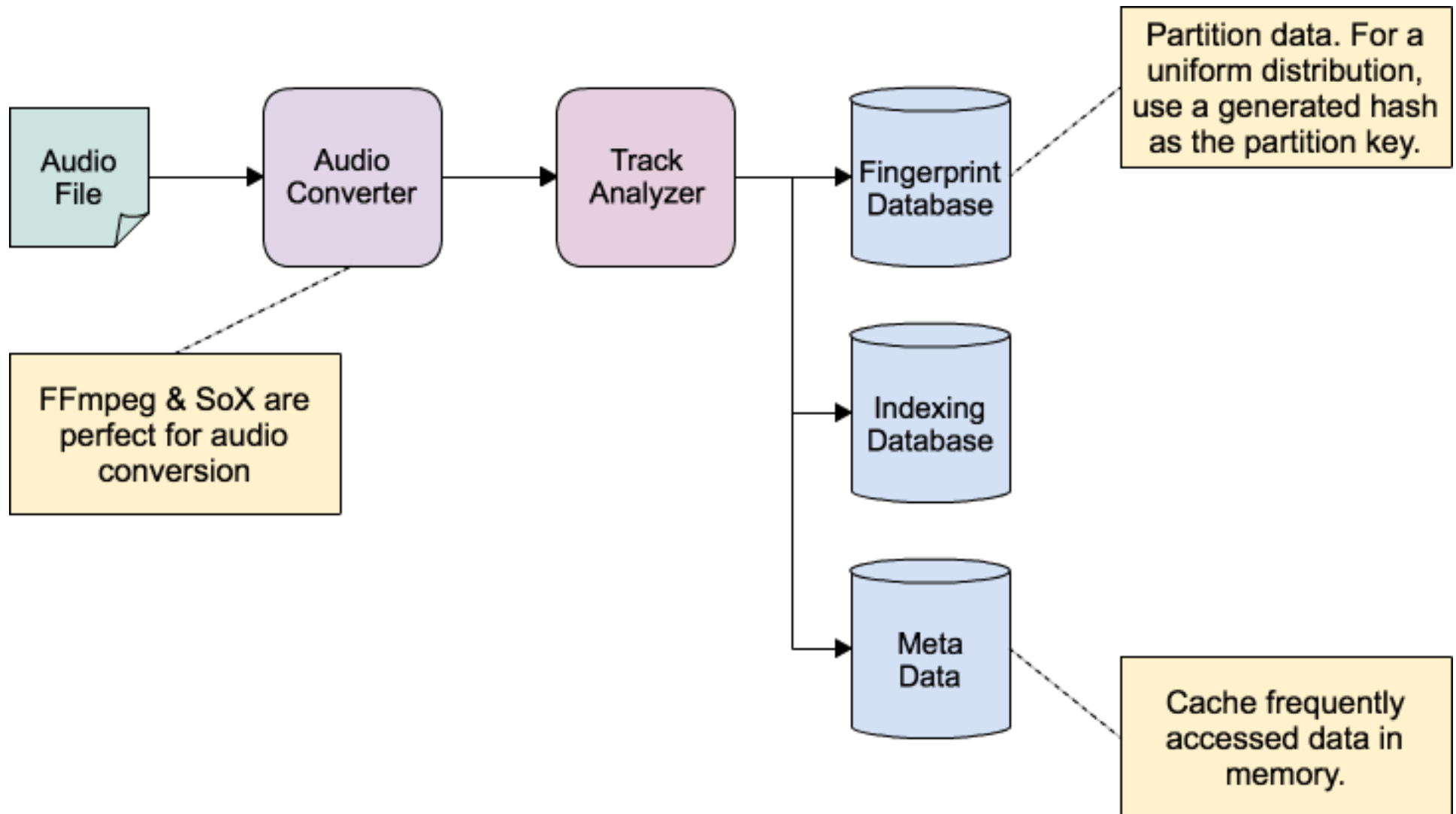
Selection of audio information

- Should be **robust**
- Should be as **unique** as possible

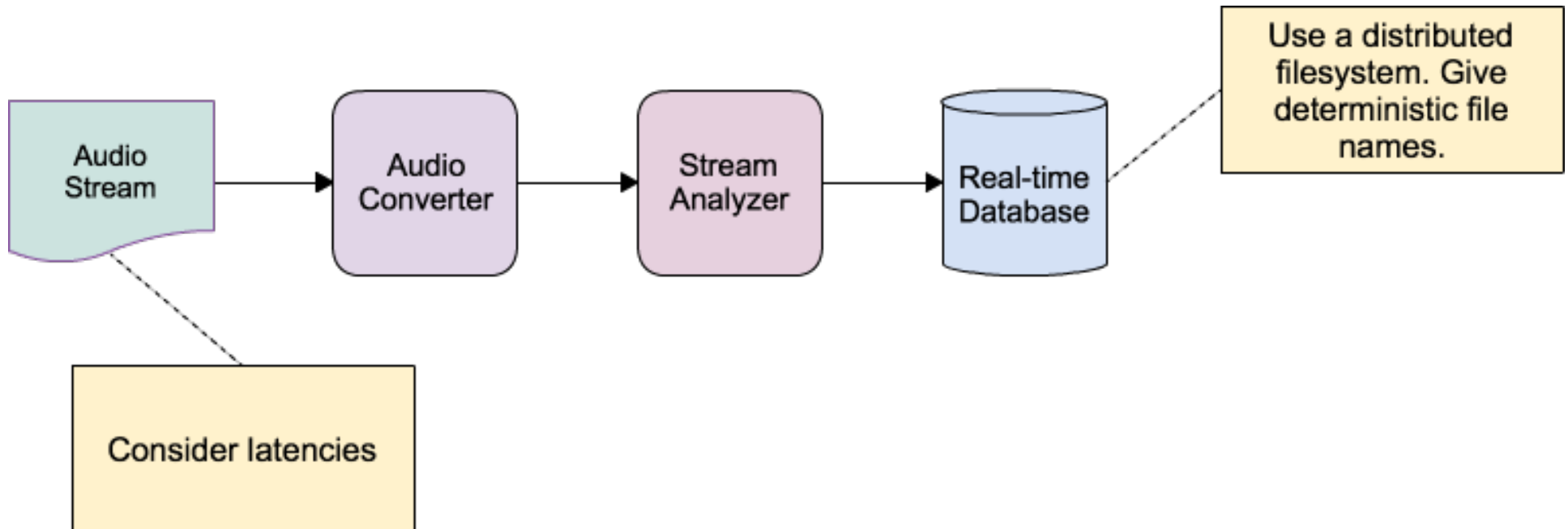
The FFT algorithm

- Managing losses due to the uncertainty principle
 - $\text{Time-resolution} = 1 / \text{Frequency-resolution}$
- Discrete-time FT or Short-time FT
- # of FFT points

Static Database



Streaming Database



Streaming Database

Stream name

Timestamp

In YYYYMMDDHHAB format

A: {0, 1, 2, 3, 4, 5} → High minute

B: {0, 2, 4, 6, 8} → Low minute

FOSDEM / 201601301648.fingerprint

Content: $T = \text{YYYYMMDDHHAB}$ file contains fingerprints from the moment T to $T + 4$ minutes

Reading: At $t = \text{YYYYMMDDHHAB}$ moment, the file corresponding to the
 $T = t - 2 - (B \& 1)$
timestamp will be opened.

Writing: At $t = \text{YYYYMMDDHHAB}$ moment, files corresponding to
 $T1 = t - 2 - (B \& 1)$
 $T2 = T1 + 2$
timestamps will be written.

Identification

Find the best matching fingerprint, if there is any

Strategy

- Reduce the search space by elimination
- Rank candidates by detailed comparison

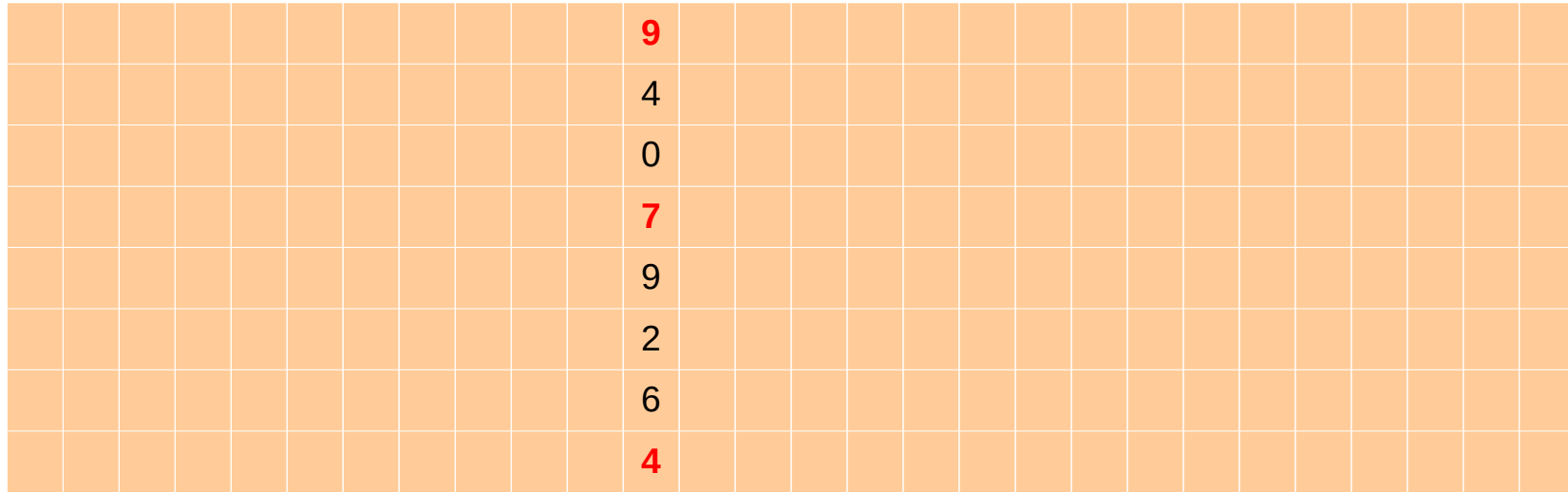
Outcomes

- **True positive:** We found the correct match
- **True negative:** We found a correct non-match
- **False negative:** We couldn't find the correct match
- **False positive:** We found an incorrect match

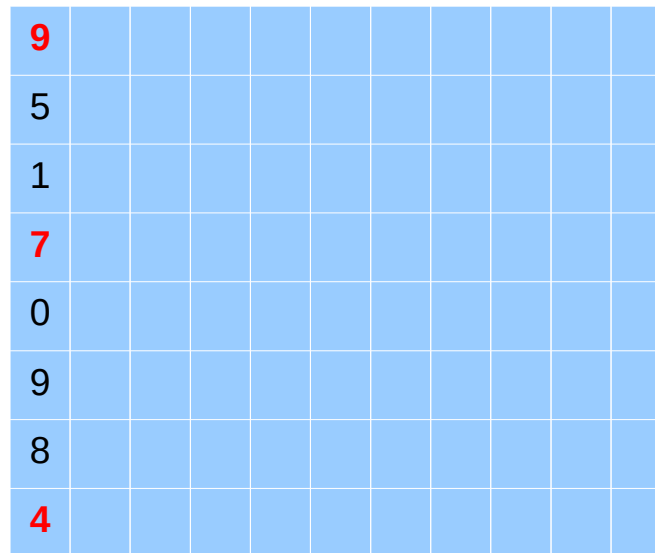
Identification: Elimination

- For each hash, try to **find exact matches**.
- For each matching hash, **calculate the time difference**.
- Create a **histogram** for time difference vs. match count.
- Eliminate candidates where the best histogram **score is less than** a predefined value.

Identification: Ranking



Spectrum score: 3
Window score: 106



Shift the window



Testing & Optimization

- Mix samples with:
 - White noise of varying volumes
 - Pre-recorded noise
- Record samples under different acoustic conditions
- Make the configuration dynamic and use a machine learning algorithm to select the best configuration

THANKS!

More will be at:

github.com/wizard/fosdem2016

- Links to **open-source software**
- **Source code** for everything we talked about
- Markdown **documentation** for this presentation
- Dockerfile

References

- FOSDEM icon: <https://fosdem.org/2016/>
- Email icon: <https://thenounproject.com/term/mail-with-at-sign/71812/>
- FFmpeg: <https://www.ffmpeg.org/>
- SoX: <http://sox.sourceforge.net/>
- Sonic Visualizer: <http://www.sonicvisualiser.org/>
- Audacity: <http://audacityteam.org/>
- PostgreSQL: <http://www.postgresql.org/>
- Redis: <http://redis.io/>
- Solr: <http://lucene.apache.org/solr/>