

Automatic Tuning of the OP-1 Synthesizer Using a Multi-objective Genetic Algorithm

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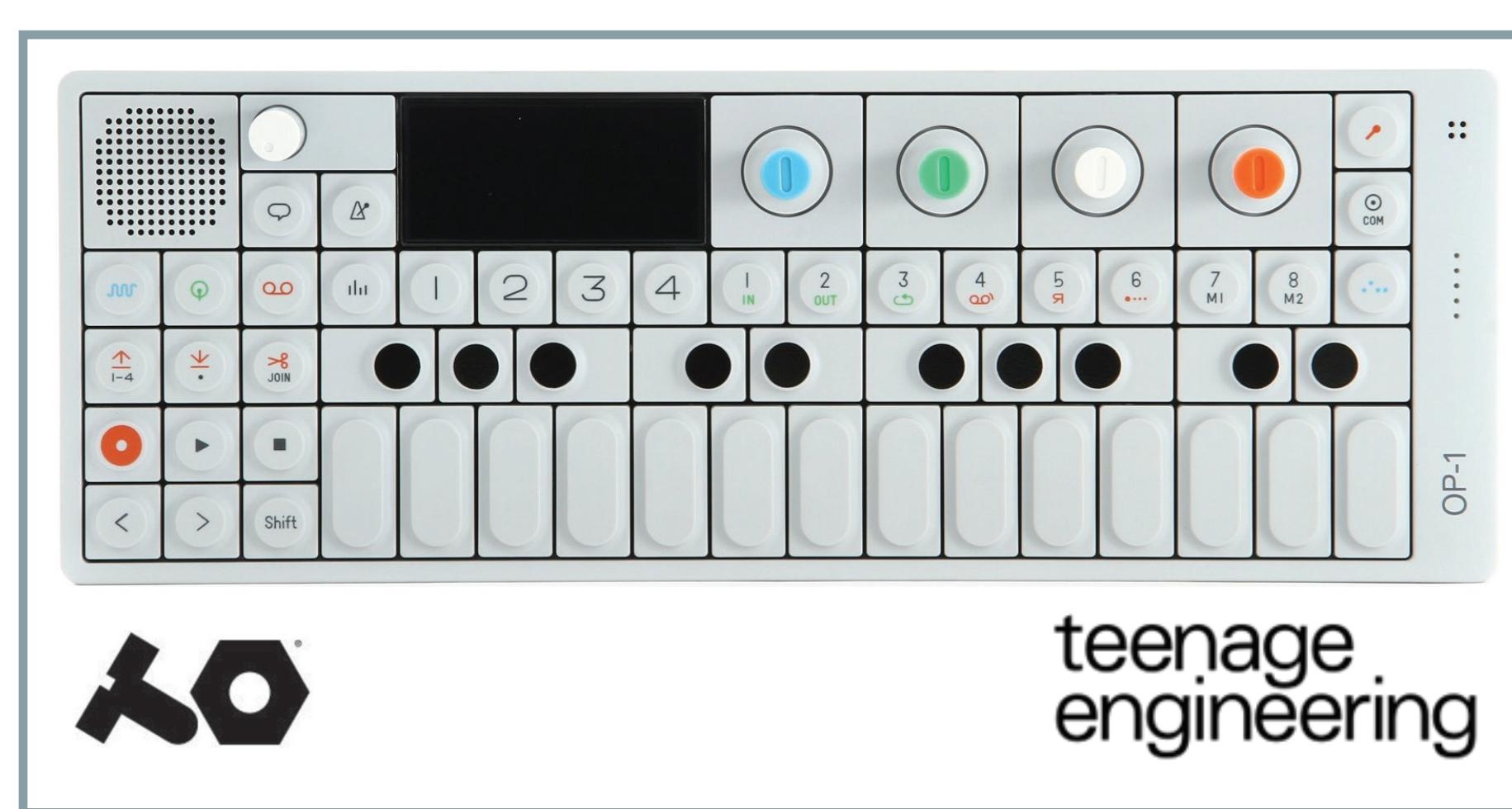
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Introduction

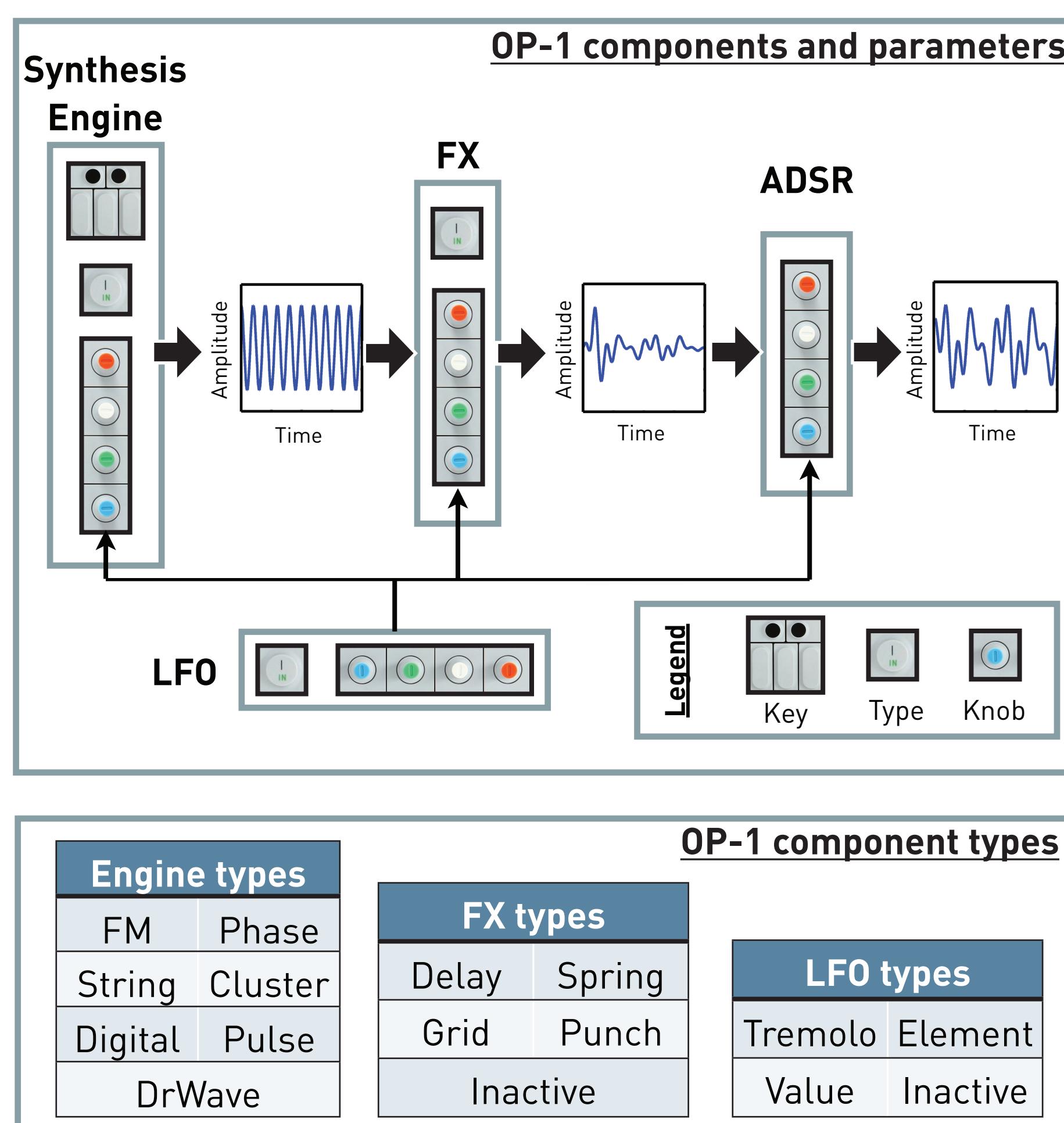
Efficient control and exploration of a synthesizer's sound space requires theoretical and/or empirical knowledge of the related synthesis technique. For a modern synthesizer, exploring large parameter spaces can become a tedious and time-consuming activity.

OP-1 Synthesizer

In this research, we use the OP-1, the all-in-one portable synthesizer, sampler and controller developed by Teenage Engineering (TE):

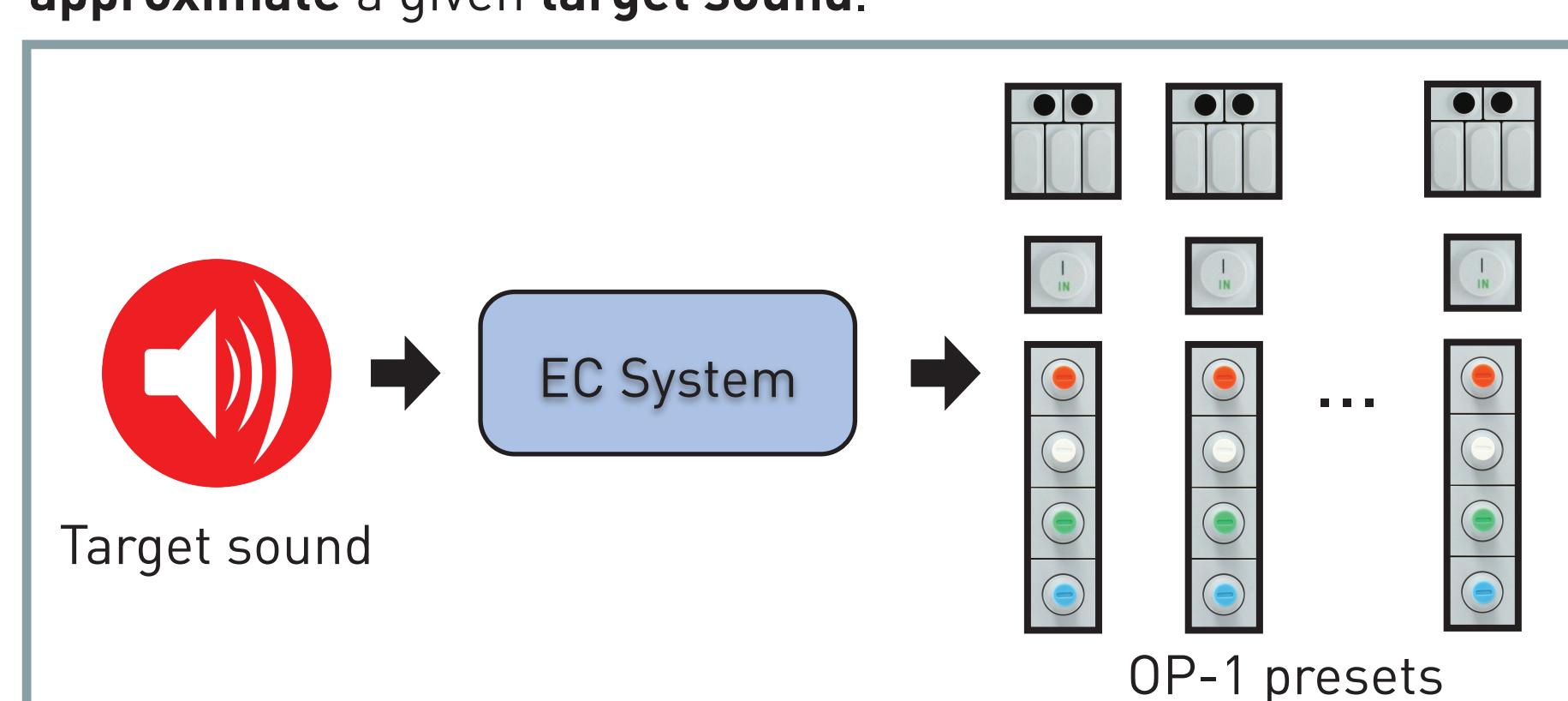


TE provided us with a library that embeds most of the functionalities of the OP-1. It includes several synthesis engines, FXs and LFOs:



Goal

In this research, we explore how **Evolutionary Computing** (EC) can be used to identify **multiple OP-1 presets** to replicate or approximate a given **target sound**.



Challenges

Searching the OP-1 parameter space to approximate a given target sound has all the characteristics of a real-world problem:

- Large search space **complexity** (10^6 possible distinct OP-1 presets),
- **Discontinuous** search space (ex: when switching from one synthesis engine to another),
- **Not fully deterministic** output.

A multi-objective approach

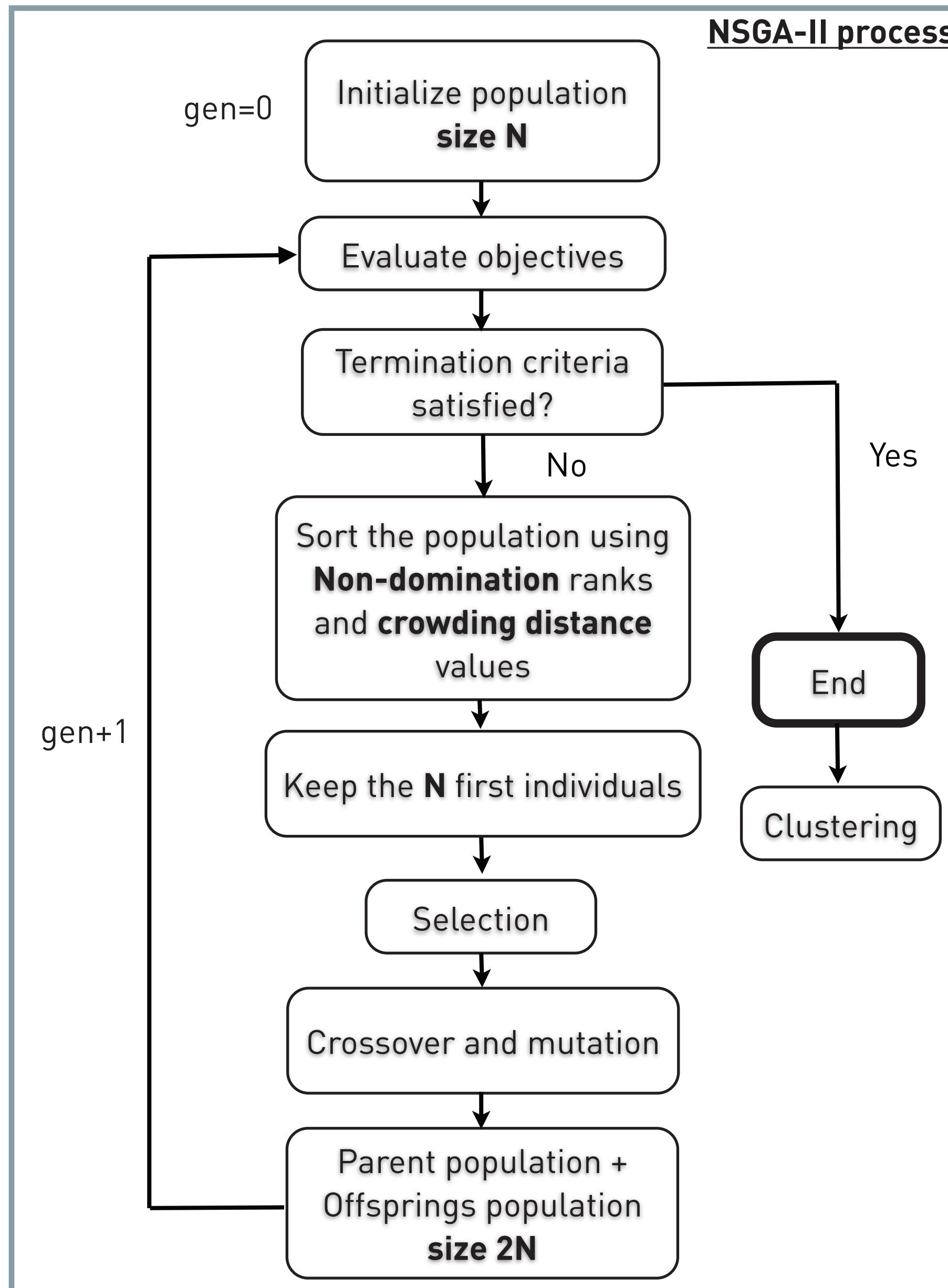
We consider **3 objective functions** to minimize in parallel. These functions are the **Euclidian distances** between the target sound and the candidate sound for the 3 following features:

- The **Fast Fourier Transform** (FFT) on the entire sound,
- The **amplitude envelope**,
- The **Short-Time Fourier Transform** (STFT).

NSGA-II

The **Non-dominated Sorting Genetic Algorithm-II** (NSGA-II) is a common technique to solve multi-objective optimization problems. The principal characteristics of this algorithm are the following:

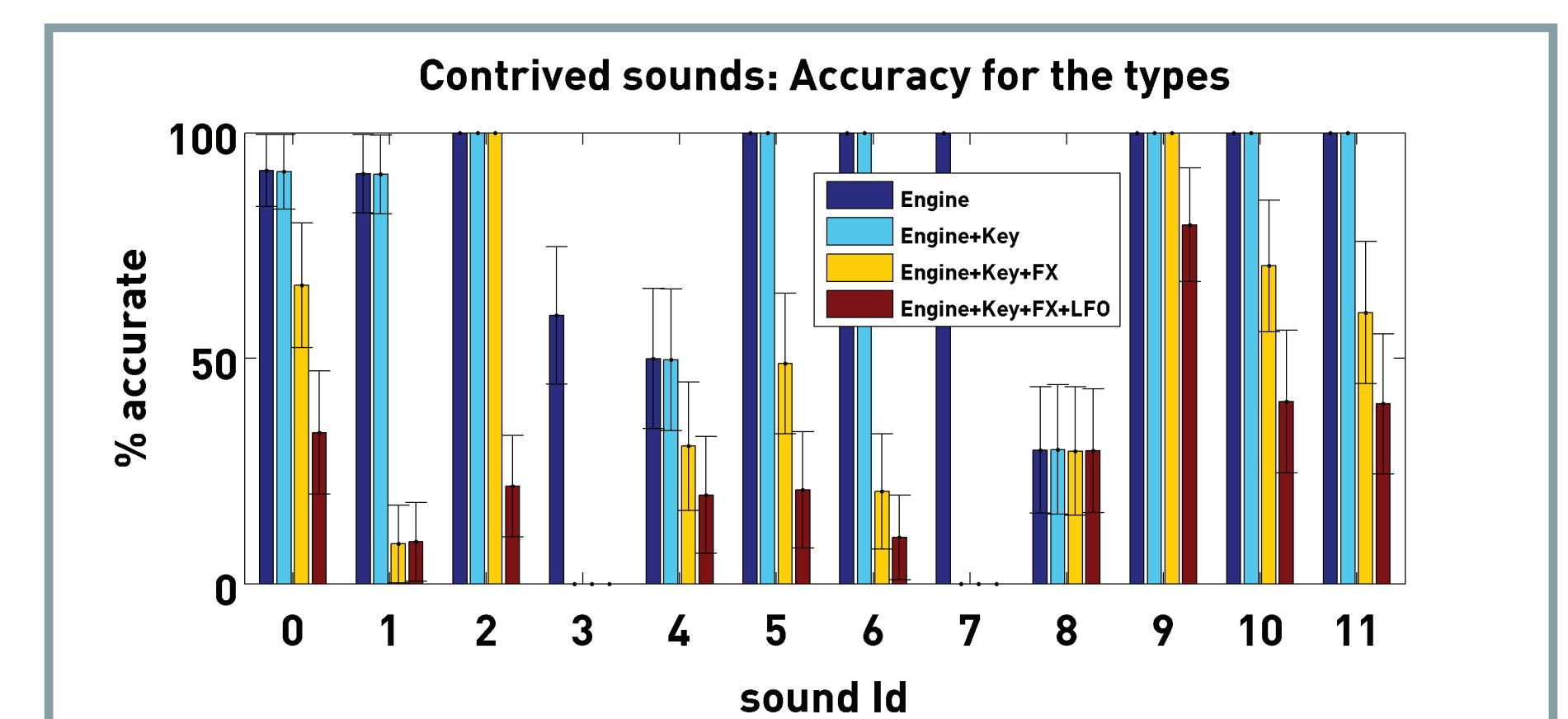
- **Elitism**
- **Non-dominated sorting**
- **Diversity preservation**



When the returned **set** of OP-1 presets is **too large** to be easily handled by an user, we implement a **k-mean clustering** method and return the **OP-1 preset** the closest to the **centroid** for each cluster.

Evaluation - Contrived sounds

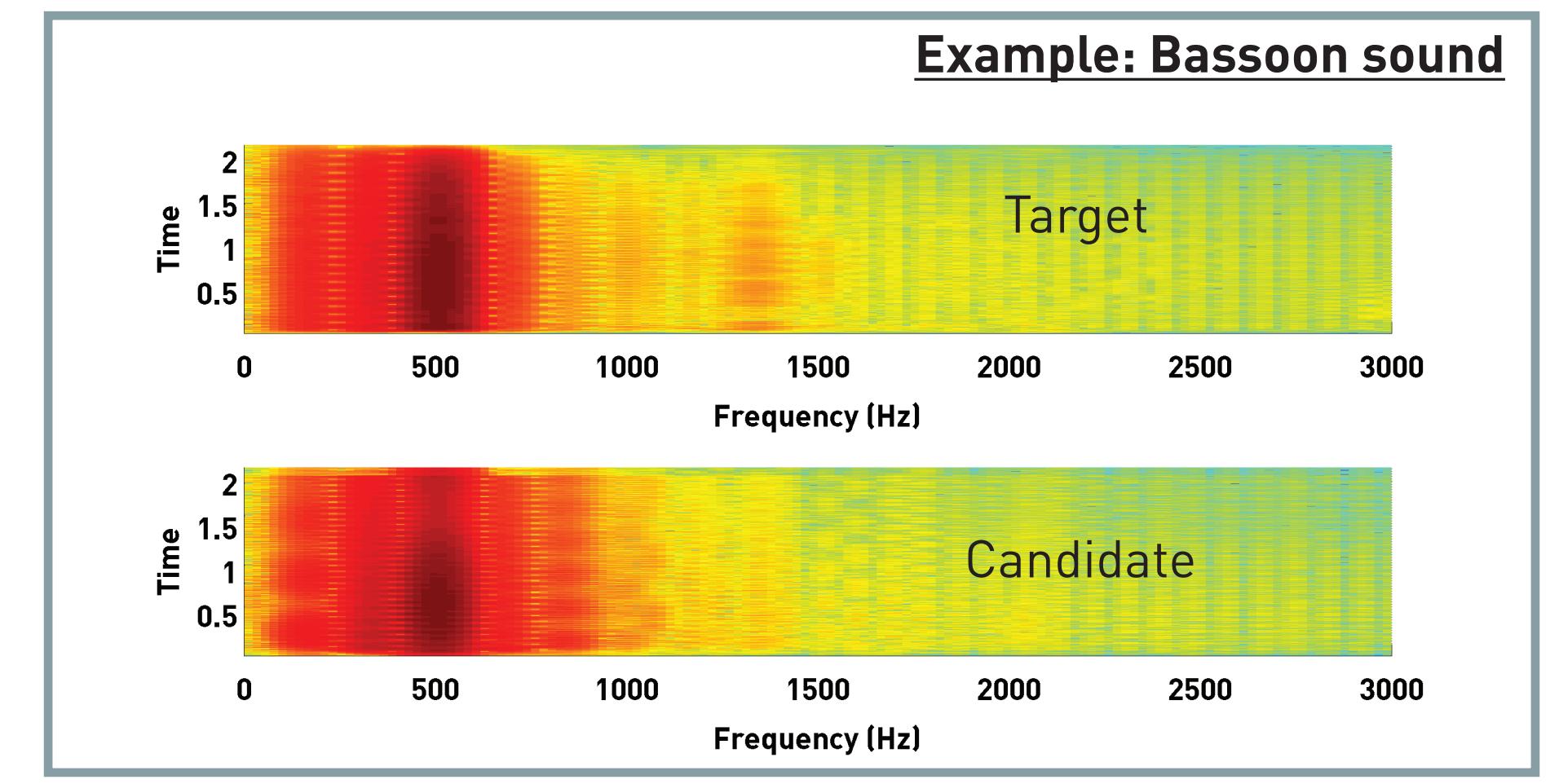
This evaluation used **OP-1 generated sounds** as target sounds for our system. This way, the **target preset is known** and we ensure that **a solution exists**.



- For most of the sounds, our system is able to successfully identify the right engine and the right key (> 90% accurate).
- The accuracy drops with the FX and LFO.
- The results are variable given the nature of the target sound (engine, FX and LFO used).

Evaluation- Non-Contrived sounds

The second evaluation used **recorded sounds** or **synthesized sounds** generated by a synthesizer other than the OP-1.



- Candidate sounds **perceptually** similar to the targets.
- **Spectra and envelopes** similar to the targets.
- Performance dissimilarities given the nature of the target.

Conclusions

- A **Non-dominated Sorting Genetic Algorithm-II** (NSGA-II) is presented that enables the user to receive **a set of solutions** rather than a unique solution as with previous systems.
- A **3 objective fitness function** including **FFT**, **Envelope** and **STFT** is developed which addresses some of the difficulties associated with the exploration of a **multi-modal search space** such as the OP-1 parameters space.
- A **k-mean clustering** method has been developed to better analyze and explore the set of final solutions.
- An **evaluation** is proposed using **contrived** and **non-contrived** sounds. The experiments revealed the **capabilities of our system** to optimize the parameters of the OP-1 synthesizer to approximate both kinds of target sounds.

Acknowledgments

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