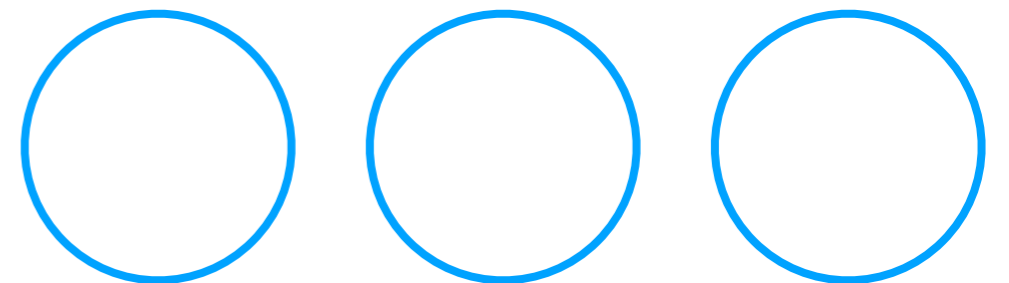


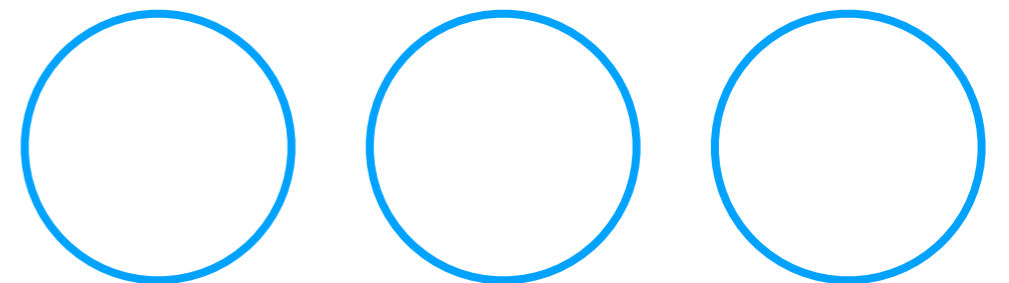
Approaches to Live Performance and Composition with Machine Learning and Music Information Retrieval Analysis

Ted Moore
CHIMEFest 2019



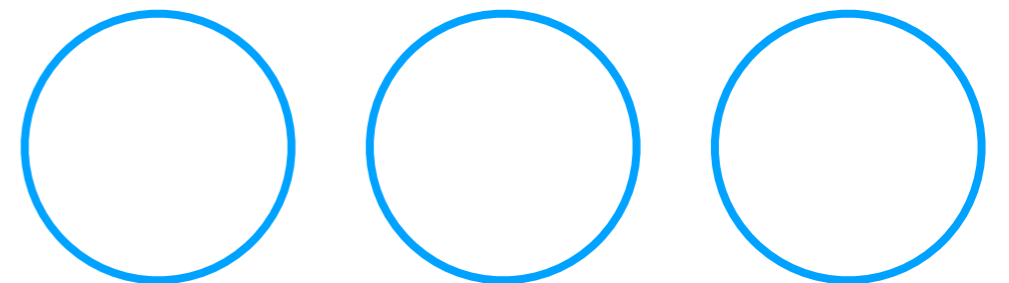
My Practice

- Composer (electronics + acoustic instruments)
- Improviser (electronics w/ acoustic collaborators)
- Coder (SuperCollider, Processing, openFrameworks, Python)
- Theatrical Sound Designer



Interest in Music Information Retrieval & Machine Learning

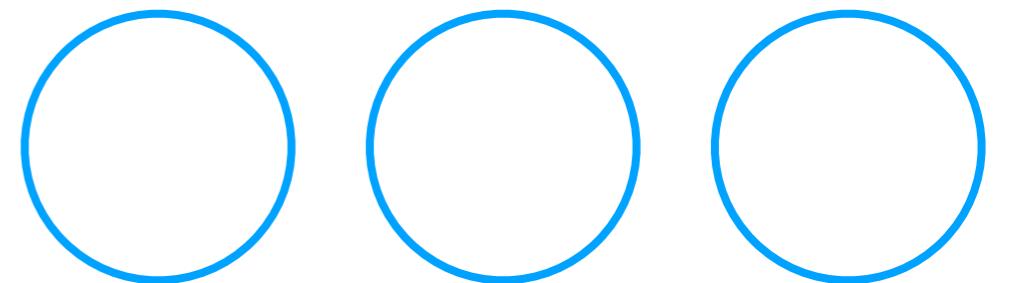
- MIR workshop at CCRMA summer 2018
- In what new ways can I approach sound?
- What can an algorithm do for (with) me? What can it tell me?
- Computational thinking
- What other routes are there to the same goal?



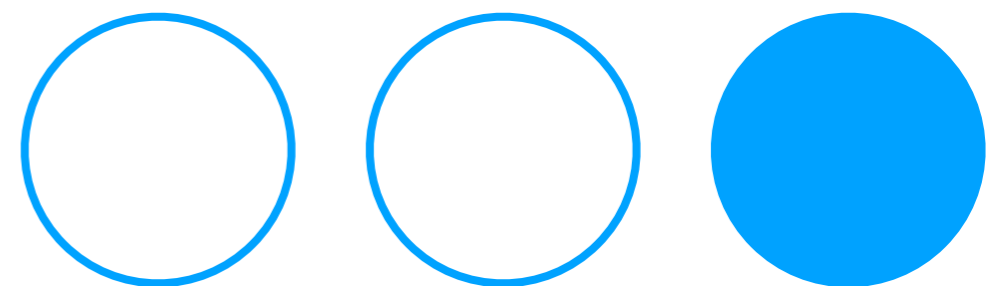
What is the goal?

Make sounds and forms that I find artistically compelling.

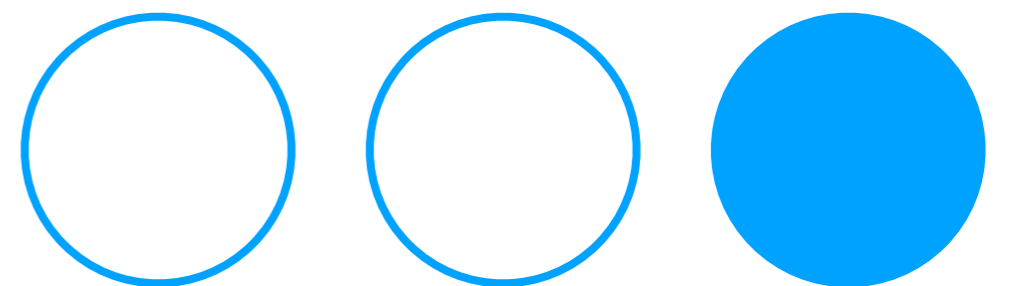
Today I share 3 examples of using these tools in that pursuit.



1. Gestural Control in MIR Space



playing random grains from a
collection of samples



Music Information Retrieval Class in SuperCollider

- 23 Dimensions of Analysis
- Onset Detection
- NRT Analysis of Files and Corpus
- Live Analysis along same Dimensions
- Returns MIRAnalysisFile, an object of its own

0: amplitude

1: fftCrest

2: fftSlope

3: fftSpread

4: loudness

5: sensoryDissonance

6: specCentroid

7: specFlatness

8: specPcile

9: zeroCrossing

10: mfcc01

11: mfcc02

12: mfcc03

13: mfcc04

14: mfcc05

15: mfcc06

16: mfcc07

17: mfcc08

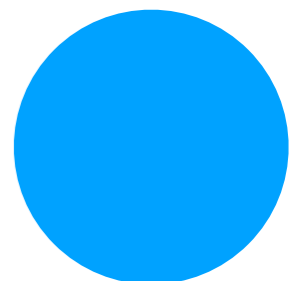
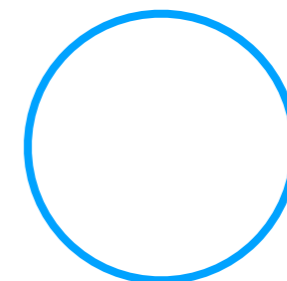
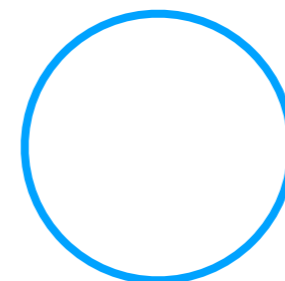
18: mfcc09

19: mfcc10

20: mfcc11

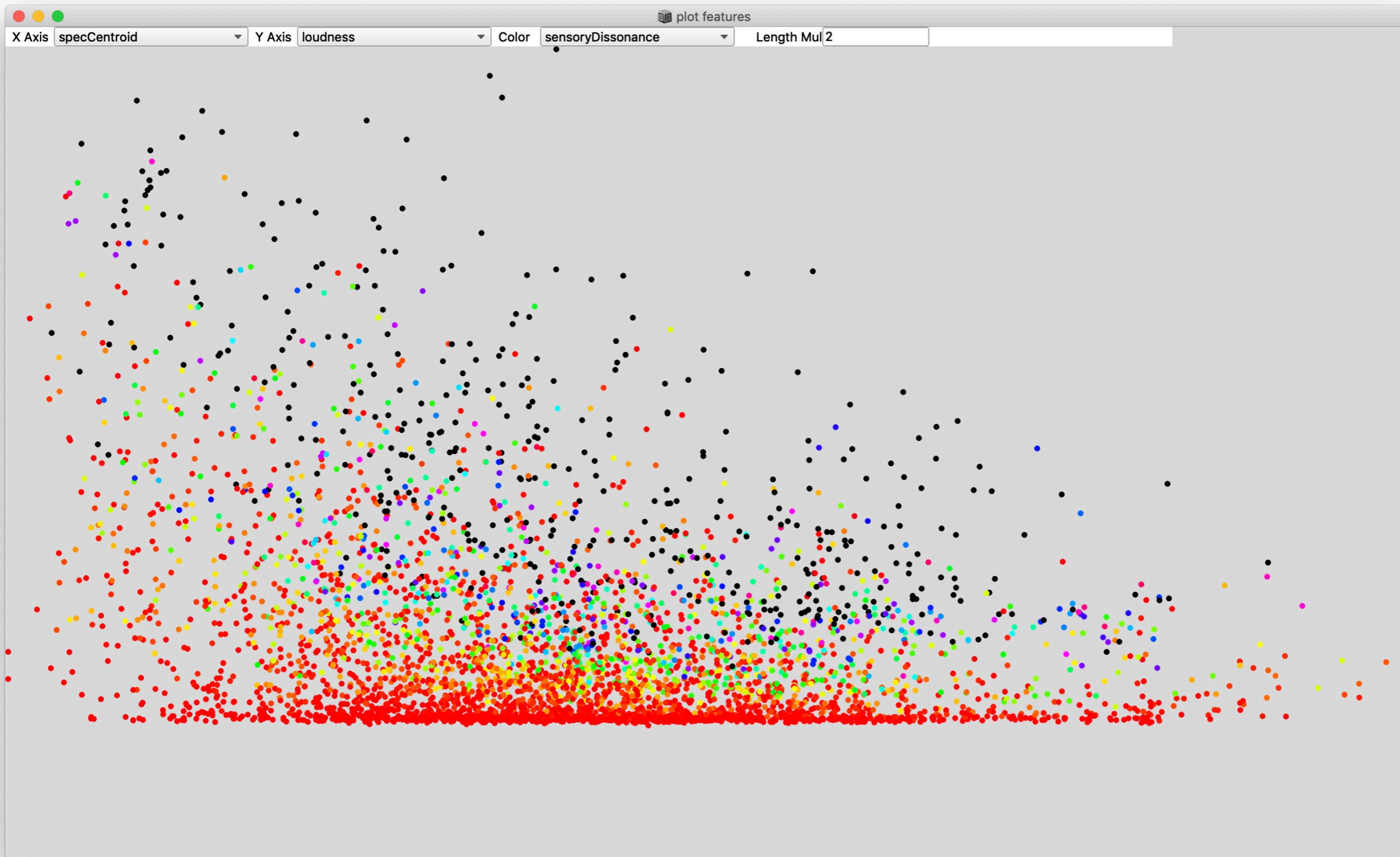
21: mfcc12

22: mfcc13



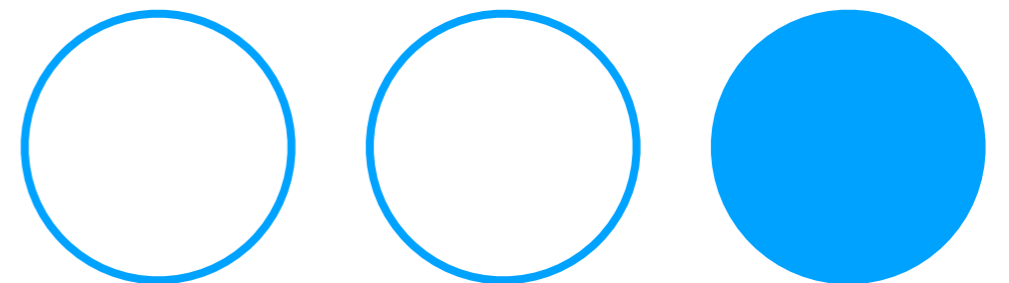
MIRCorpus class in SuperCollider

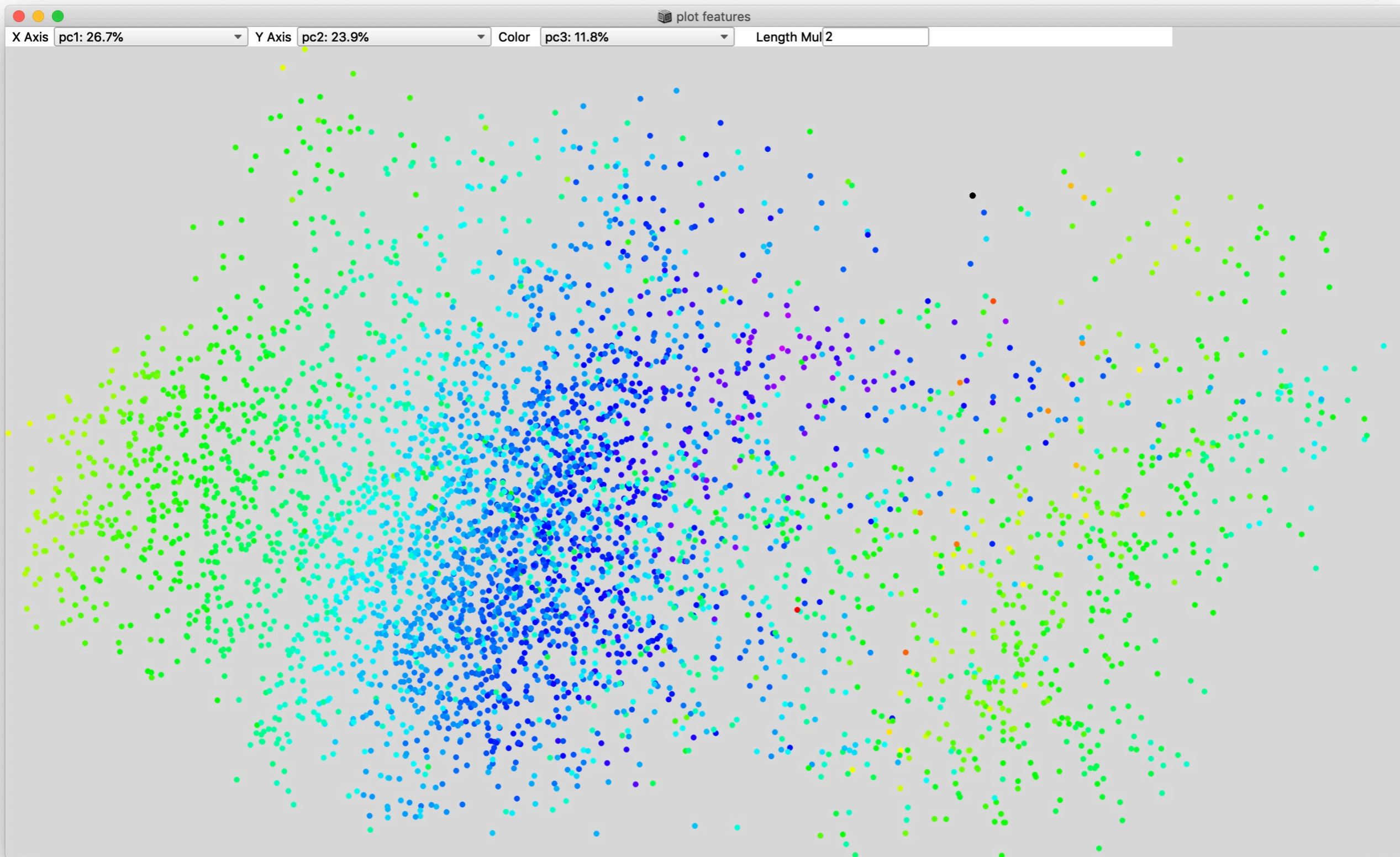
- MIRAnalysis on source corpus
- NRT
- Returns MIRCorpusItem



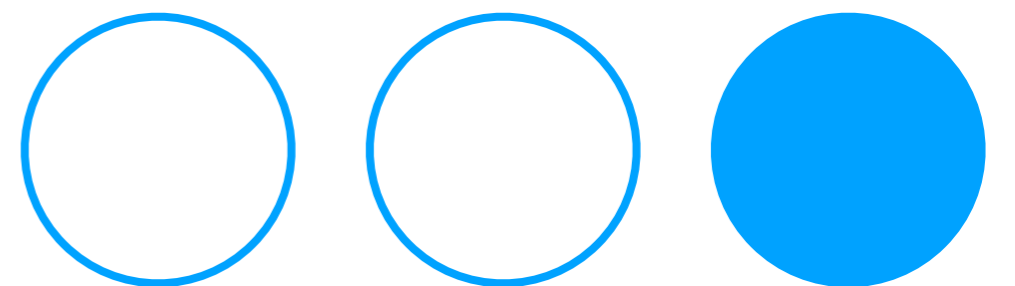
Principal Component Analysis

- Reduce the number of dimensions in a data set
- Maintain the variance in the data set
- Remove redundancy

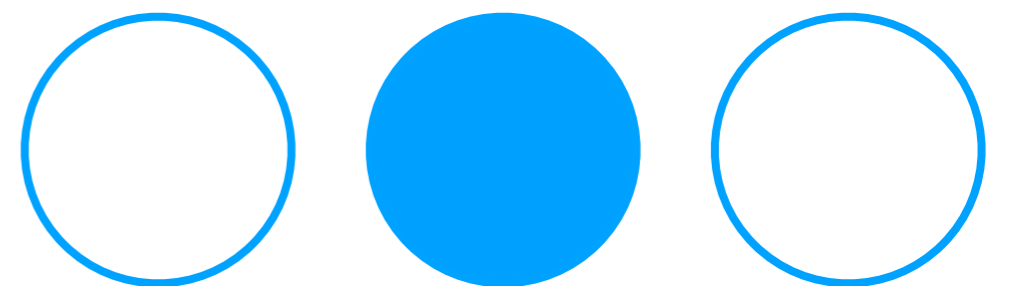




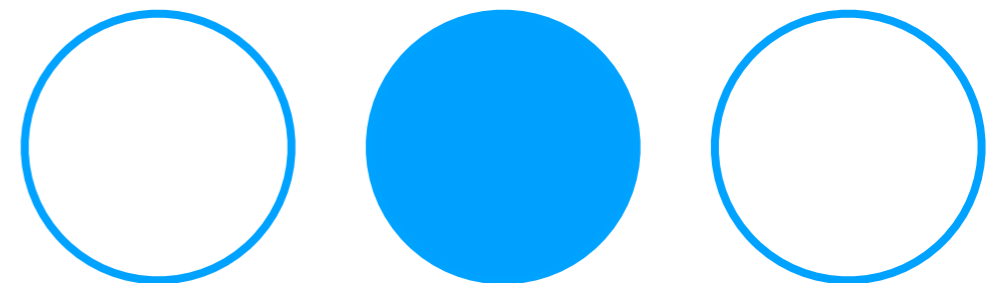
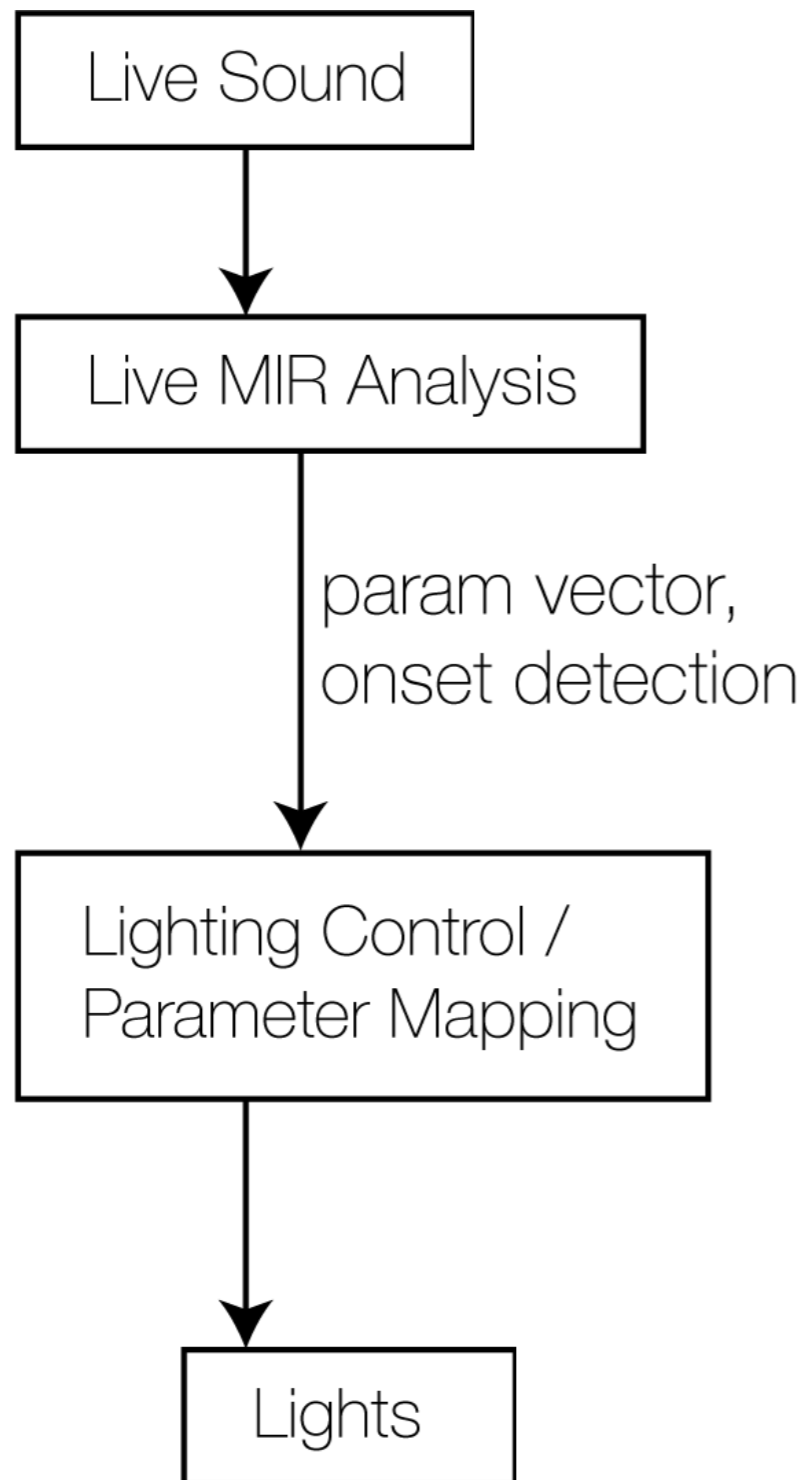
demo time



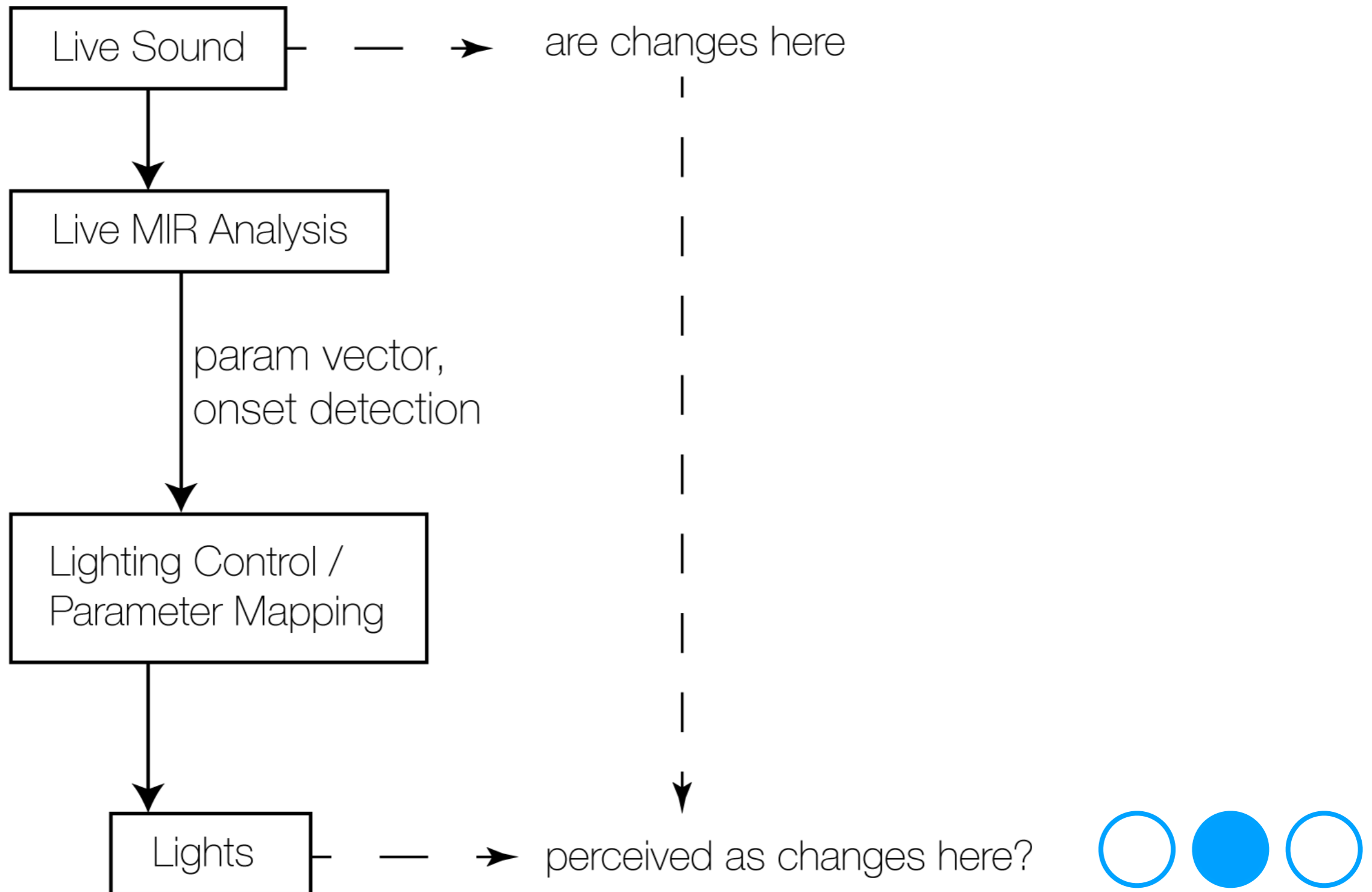
2. Live Sound Classification



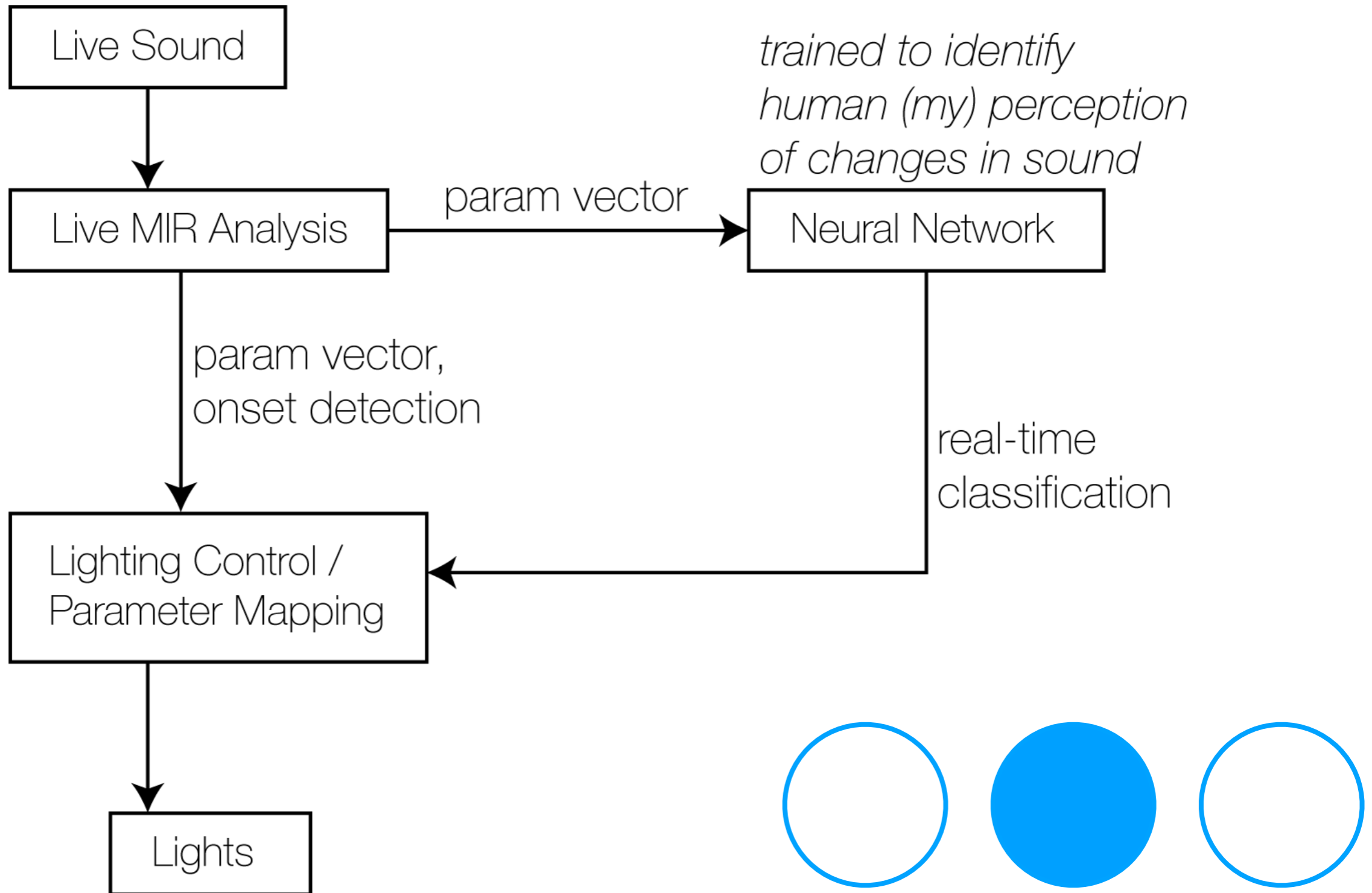
Machine Listening System



Machine Listening System



Machine Learning System





distorted_noise



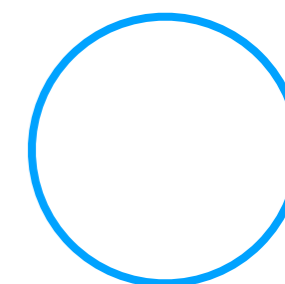
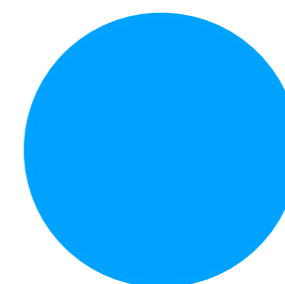
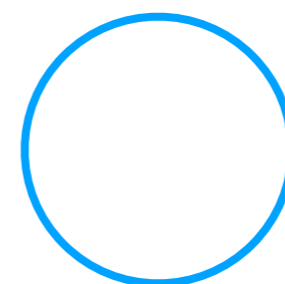
high_squeal



low_impulses



sus_noise_quiet



```

1 NeuralNetwork {
2     var <>net, <>learningRate, e = 2.71828, <shape, <>activation, <>normalizedRanges;
3
4     *new {
5         arg shape, learningRate = 0.05, activation = "relu", normalizedRanges;
6         ^super.new.init(shape, learningRate, activation, normalizedRanges);
7     }
8
9     init {
10        arg shape_, learningRate_ = 0.05, activation_ = "relu", normalizedRanges_;
11        shape = shape_;
12        activation = activation_;
13        learningRate = learningRate_;
14        normalizedRanges = normalizedRanges_;
15
16        net = shape.collect({
17            arg nNeurons, i;
18            var data = (
19                vals:Array.fill(nNeurons, {0}),
20            );
21            if(i > 0, {
22                // not input layer;
23                data.biases = Array.fill(nNeurons, {rrand(-1.0, 1.0)});
24                data.weights = Array.fill(shape[i], {
25                    Array.fill(shape[i-1], {rrand(-1.0, 1.0)});
26                });
27            });
28            data;
29        });

```



```
feedLights = FeedLightMaster([
```

```
    // distorted noise
```

```
    FeedLightMode(nLights, [
```

```
        FeedLightGroup([
```

```
            \amplitude, \myAmp, \v, ControlSpec(0.01, 1, \exp),
```

```
            \specCentroid, ControlSpec(50, 5000, \exp), \h, ControlSpec(0.5, 0.7),
```

```
            \specFlatness, nil.asSpec, \s, ControlSpec(1, 0.3)
```

```
        ]),
```

```
        FeedLightGroup([
```

```
            \amplitude, \myAmp, \v, ControlSpec(0.01, 1, \exp),
```

```
            \specCentroid, ControlSpec(50, 5000, \exp), \h, ControlSpec(0.4, 0.6),
```

```
            \specFlatness, nil.asSpec, \s, ControlSpec(1, 0.3)
```

```
        ])
```

```
    ]),
```

```
    // high squeal
```

```
    FeedLightMode(nLights, [
```

```
        FeedLightGroup([
```

```
            \amplitude, \myAmp, \s, ControlSpec(0.5, 0.9),
```

```
            \zeroCrossing, ControlSpec(3000, 10000, \exp), \h, ControlSpec(0, 0.25),
```

```
            \constant, 1, \v, nil
```

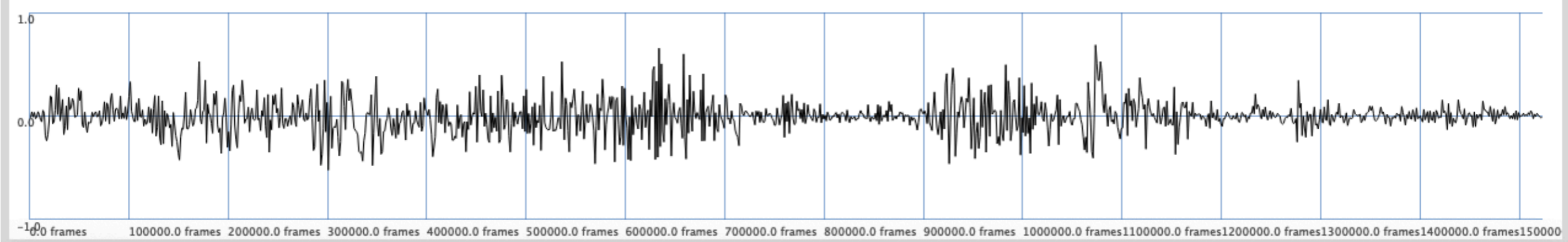
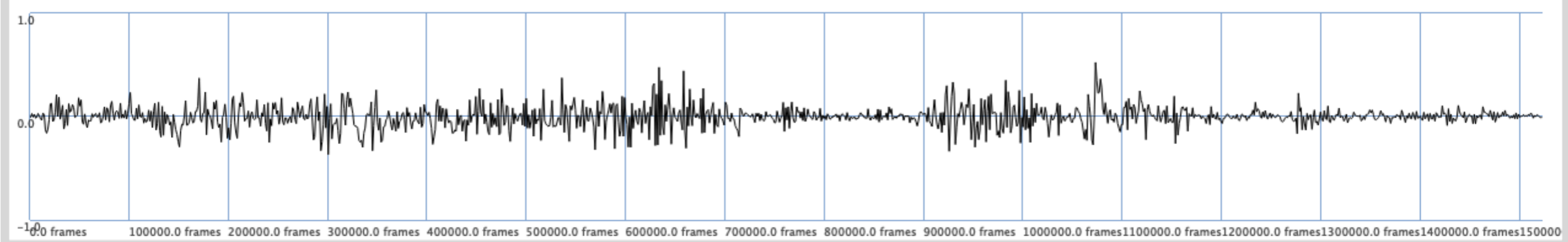
```
            // \specFlatness, nil.asSpec, \w, ControlSpec(0, 255),
```

```
            // \zeroCrossing, ControlSpec(50, 6000, \exp), \r, ControlSpec(0, 255)
```

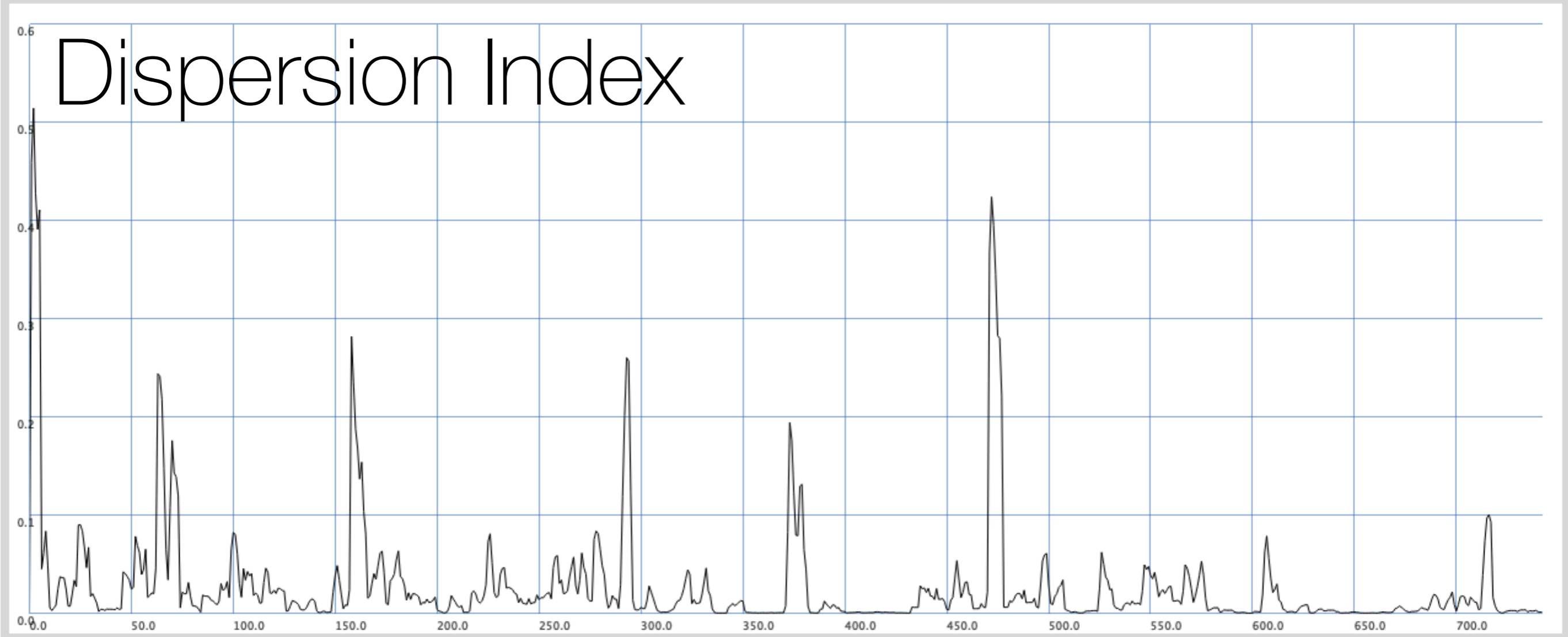
```
        ]),
```

```
        FeedLightGroup([
```

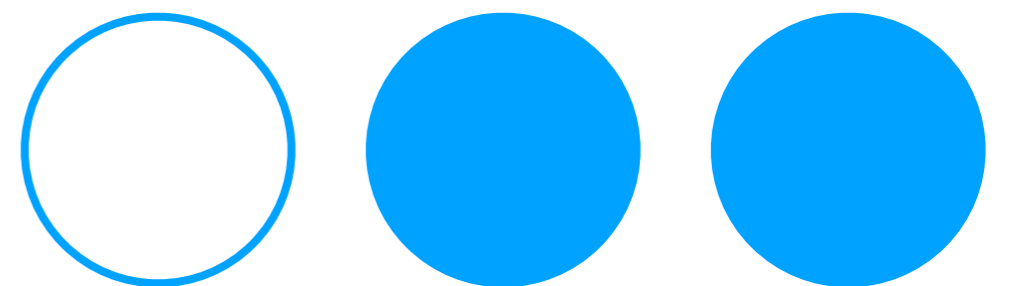
```
            \amplitude, \myAmp, \s, ControlSpec(0.5, 0.9)
```

Plot



3. Corpus Concatenation



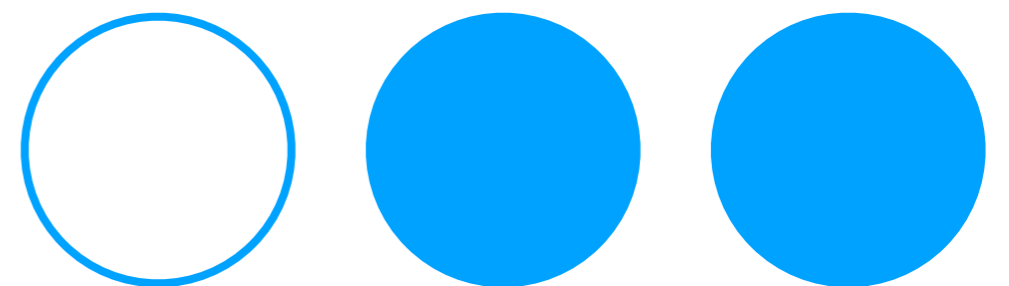
ConcatSynthNRT class in SuperCollider

Render in many ways, flexibility for rendering from different types of data

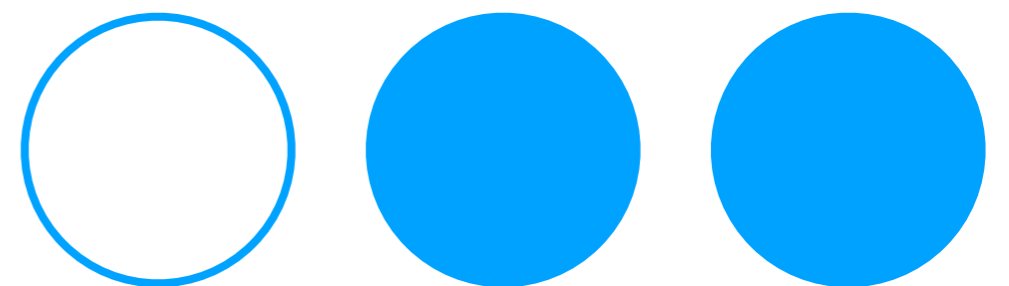
- `renderFromCorpusAndFilePath()`
 - kNN
- `renderFromCorpusAndRawFrames()`
 - kNN
- `renderFromArrayOfCorpusItems()`
 - from given path



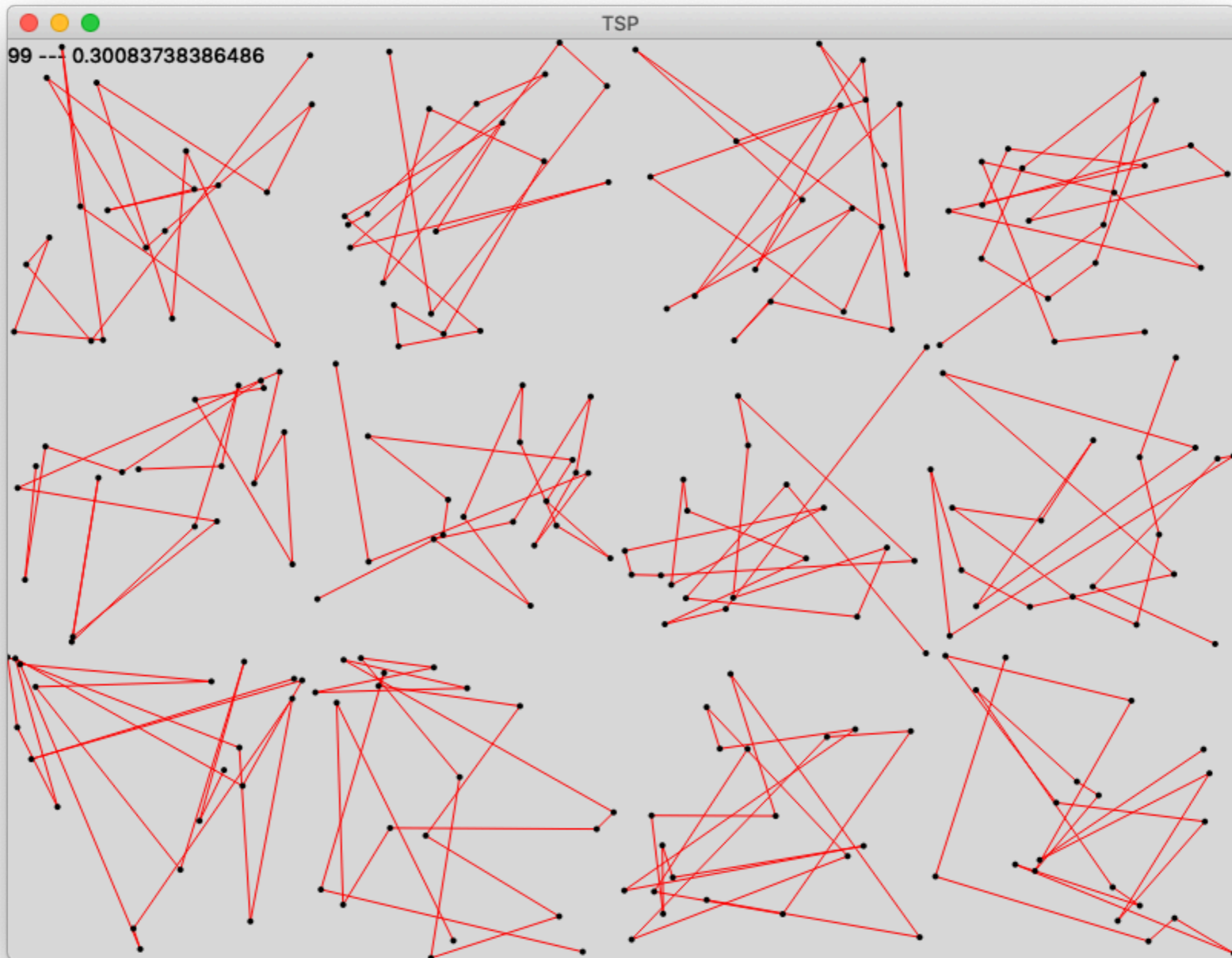
what does the fox say?



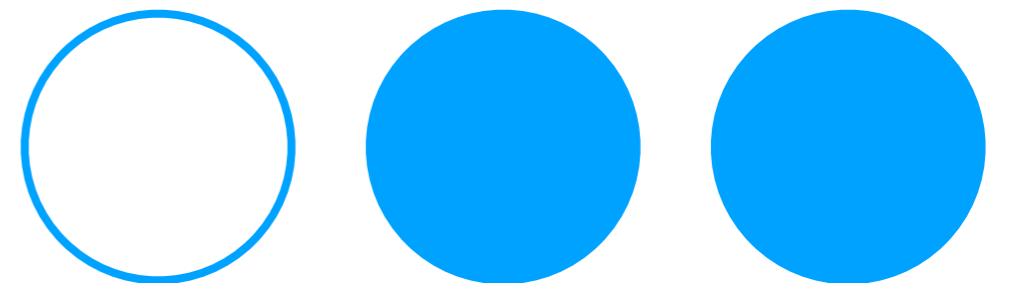
what does the fox say?

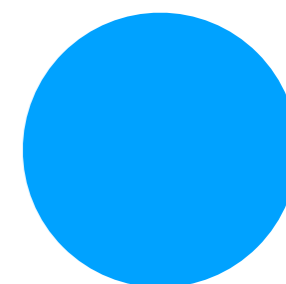
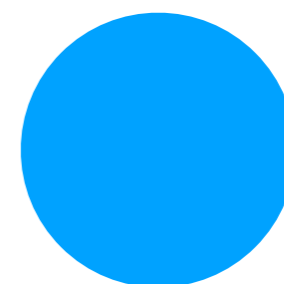
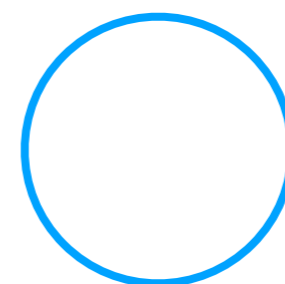
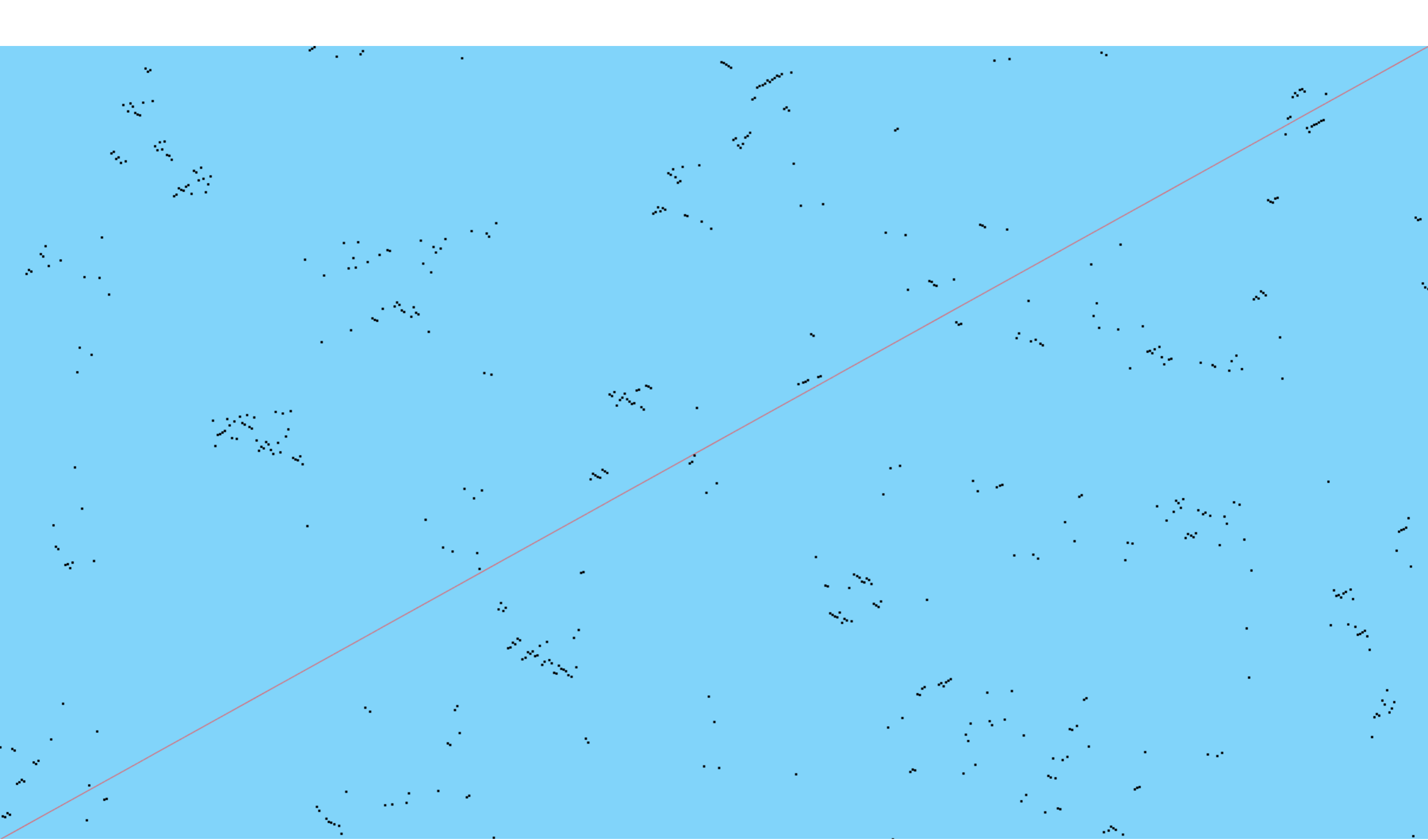


pathfinding through MIR space:
empowering algorithms to organize time

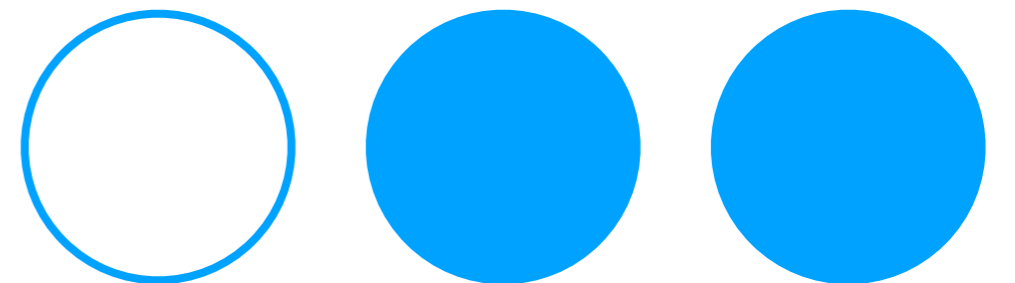


what does the sax say?





what does the corpus say?



reiny_bells 01

noisy_laptop 01

saxophone 01

eurorack 01

sustain_elec 01

drums 01

bassoon 01

no_input_mixer 01



reiny_bells 01

noisy_laptop 01

saxophone 01

eurorack 01

sustain_elec 01

drums 01

bassoon 01

no_input_mixer 01



reiny_bells 01

noisy_laptop 01

saxophone 01

eurorack 01

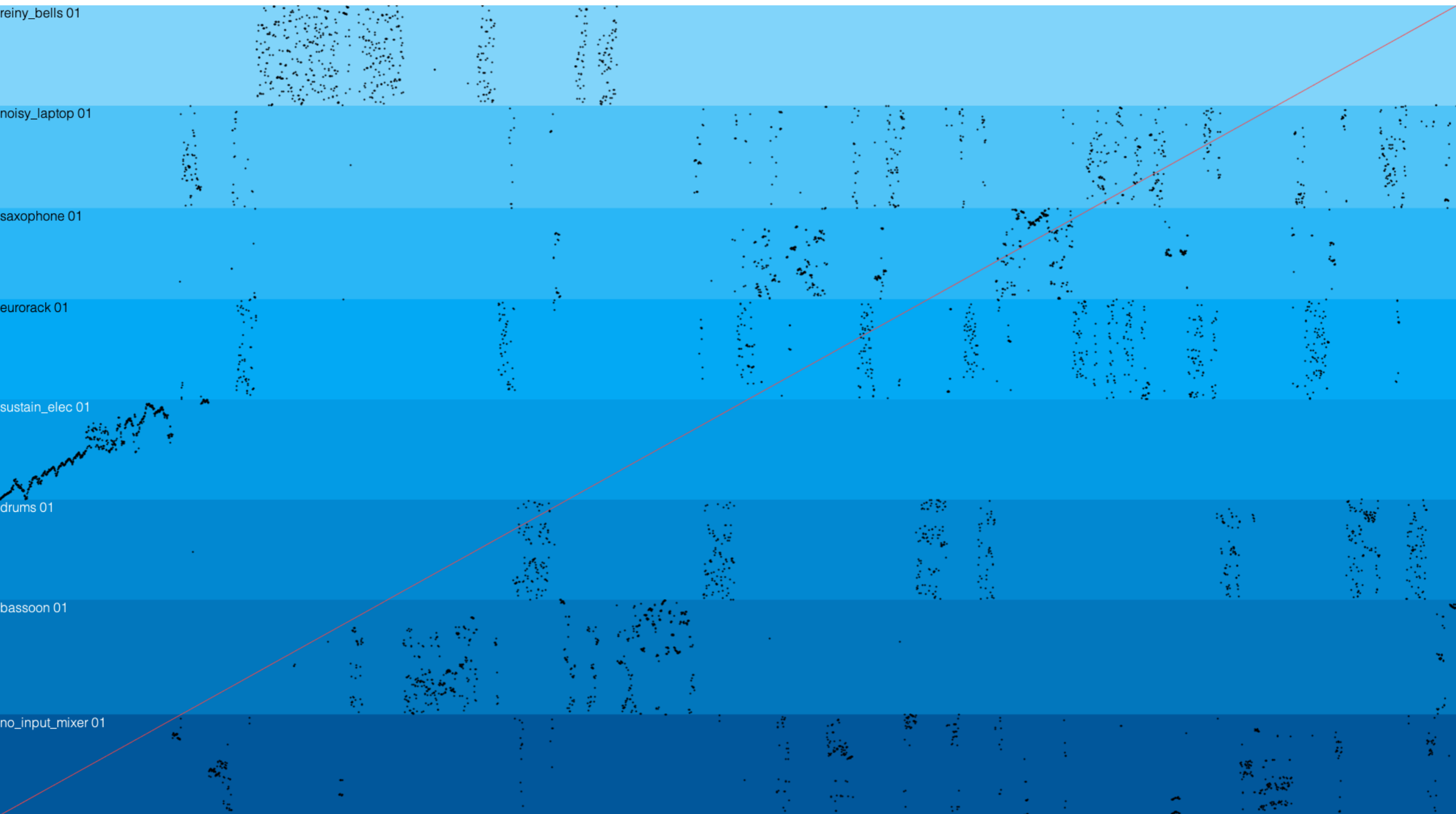
sustain_elec 01

drums 01

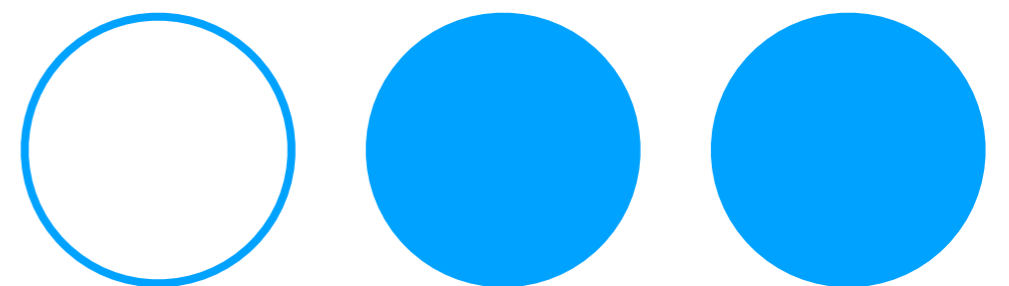
bassoon 01

no_input_mixer 01



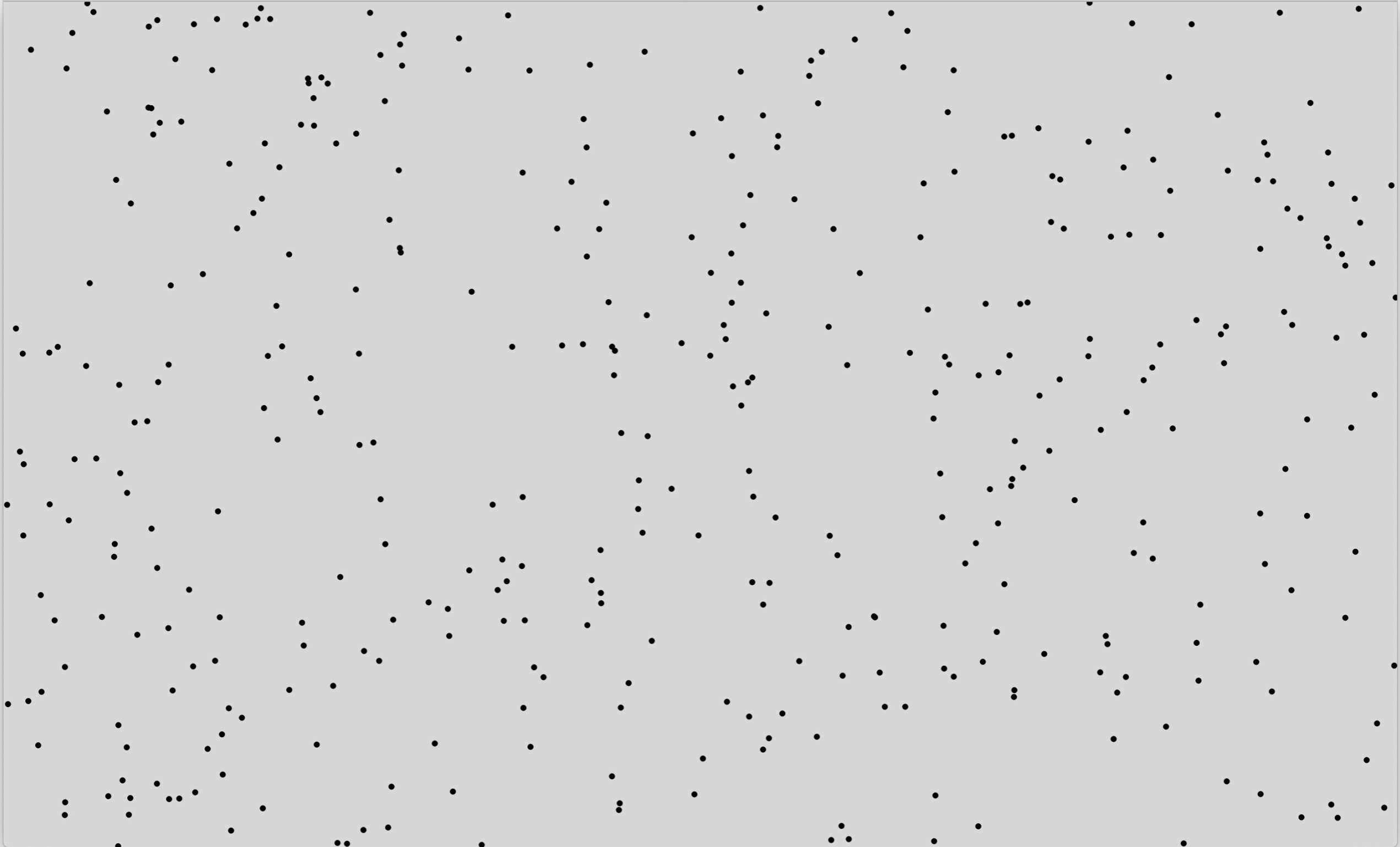


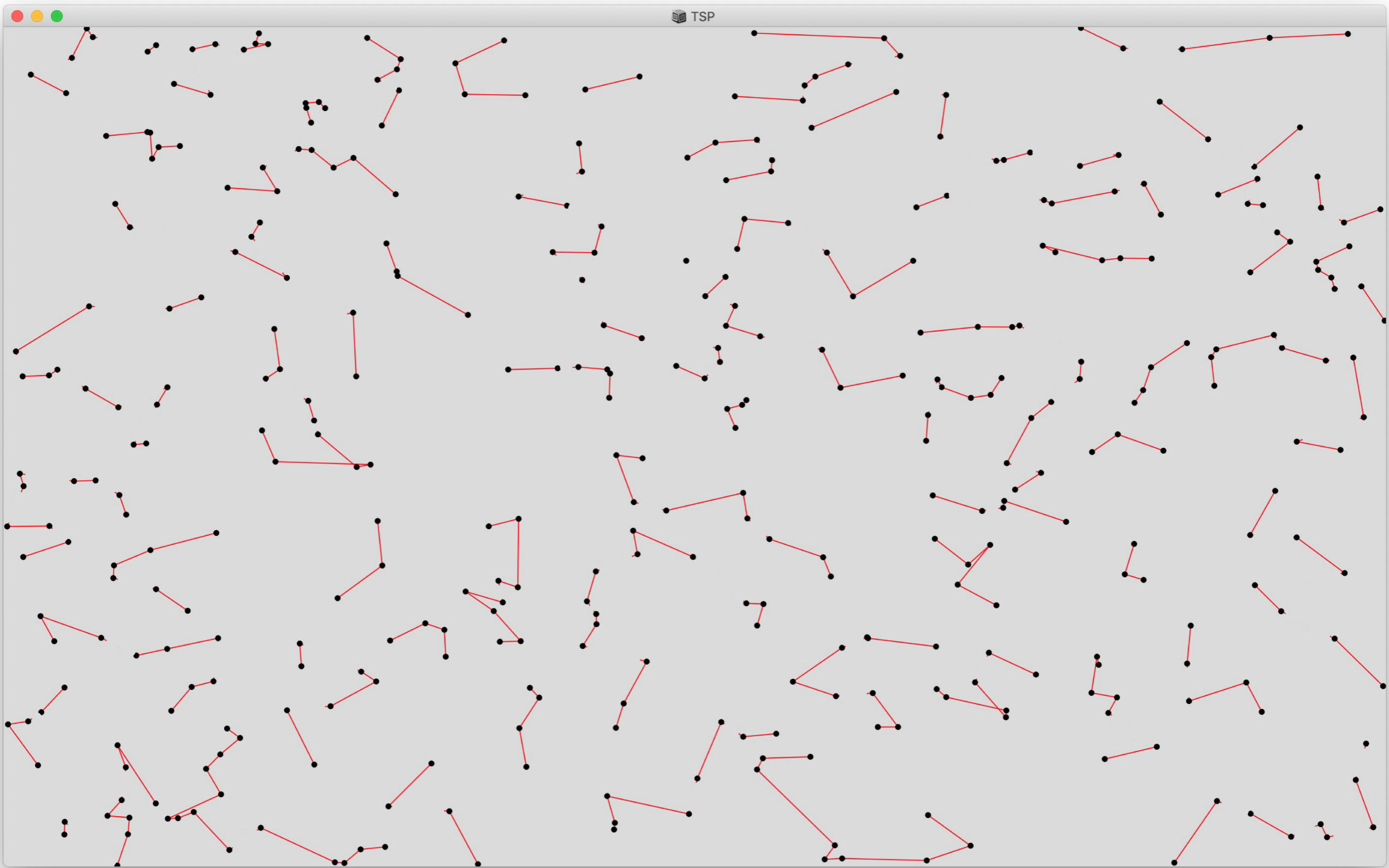
machine composed timbral
fusion, gesture, & form



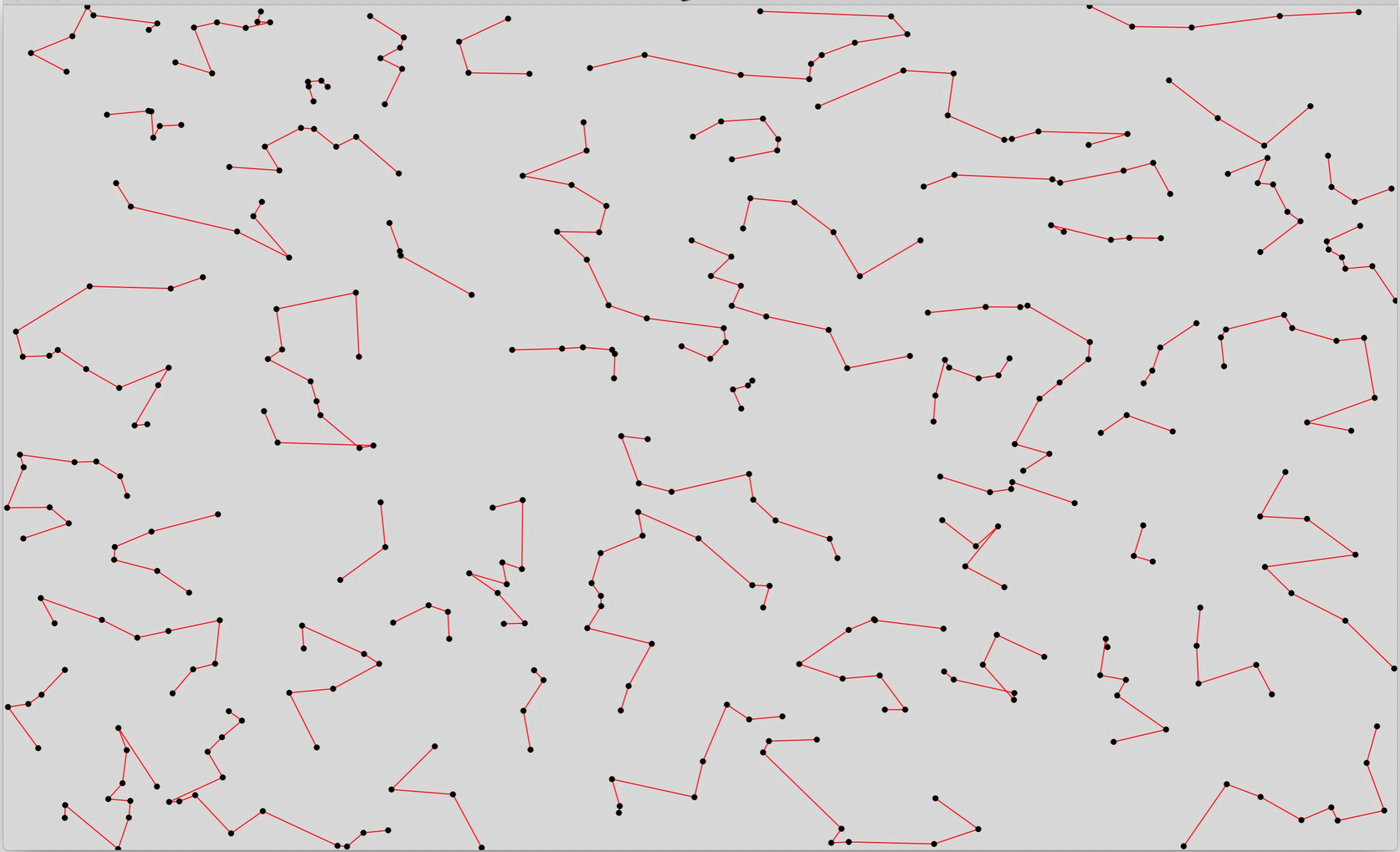


TSP

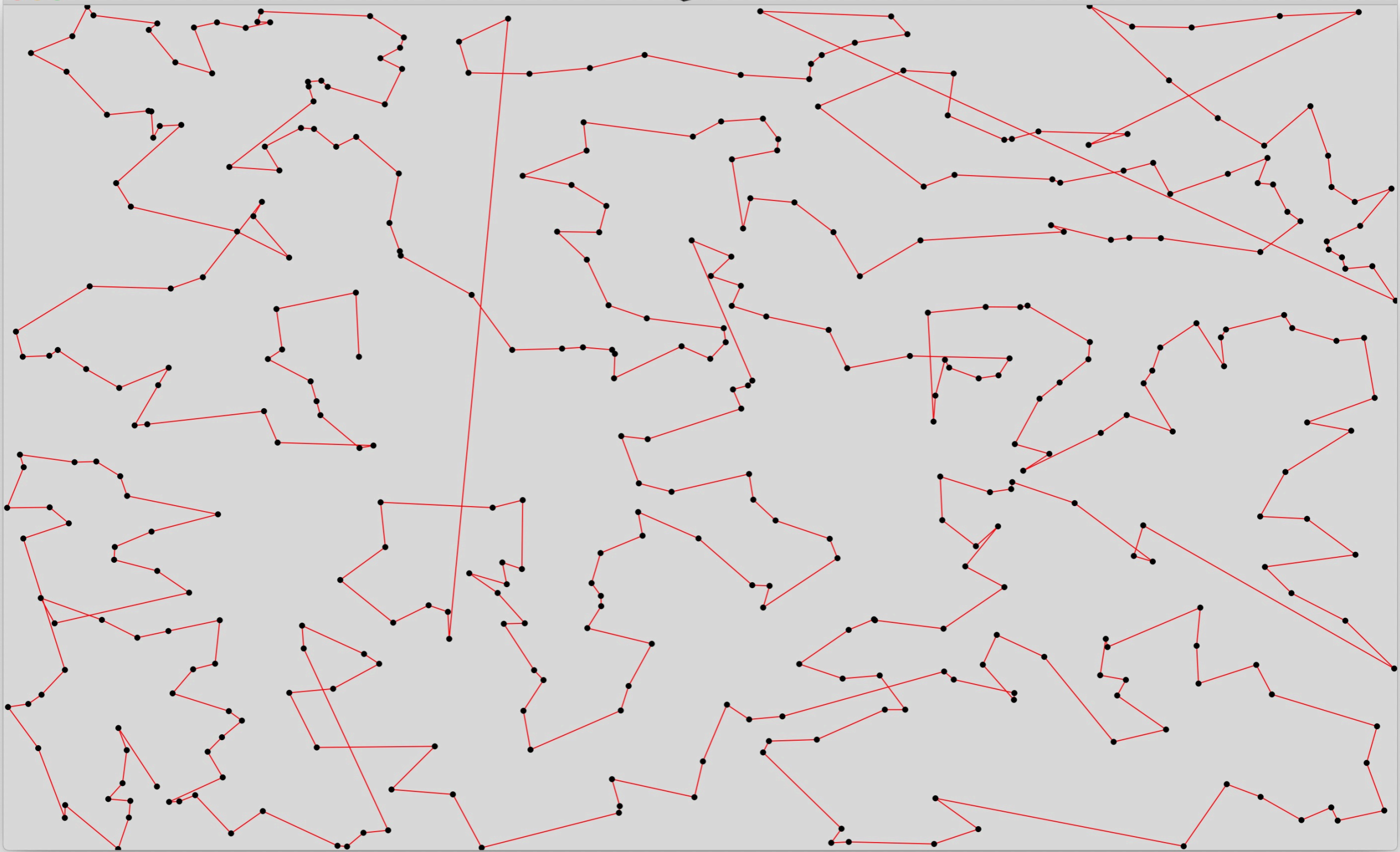




TSP



TSP



pathfinding through FFT space: towards machine learning “synthesis”

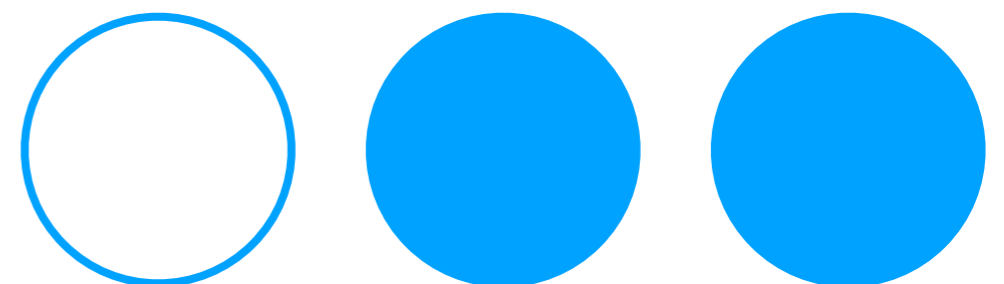
```
1 FFTNRT {
2
3     *fft {
4         arg filePath, action, fftSize = 2048, overlap = 2;
5         SoundFile.use(filePath,{
6             arg sf;
7             var data, window, frames, currentSample = 0, hopSamples, fft, imag;
8
9             hopSamples = fftSize / overlap;
10
11             data = FloatArray.newClear(sf.numFrames * sf.numChannels);
12             sf.readData(data);
13
14             // sum to mono
15             // TODO: process stereo|
16             if(sf.numChannels > 1,{
17                 data = sf.numFrames.collect({
18                     arg frameI;
19                     var val = 0;
20                     sf.numChannels.do({
21                         arg chanI;
```



pathfinding through FFT space:
towards machine learning “synthesis”

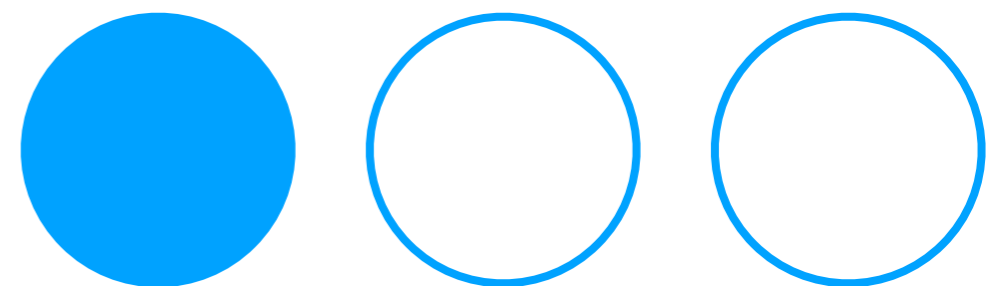
```
Python running for 08:53:33.471  
Python running for 08:53:34.471  
Python running for 08:53:35.471  
[ 6982, 6981, 4968, 8067, 8002, 800
```

saxophone, no input mixer, laptop, eurorack



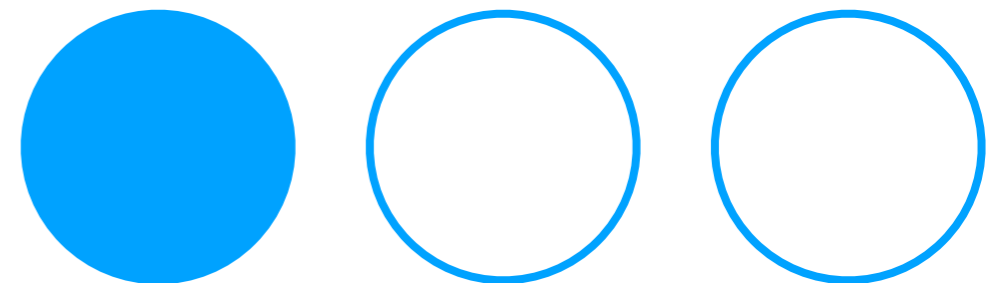
“The view according to which the novelty of a work guarantees its quality is often expressed in electroacoustic music circles, and for some it is the only criterion of worthiness.”

–Francis Dhomont, *For classicism*



Future Directions

- Introduce more sequence (time) based algorithms (RNNs, HMMs), have them “help” in real-time
- Understand lower dimensional space using VAE (instead of PCA)
- Machine Learning Synthesis (composing sequential FFT frames, GANs)
- Concatenative Synthesis not based on kNN, but on user defined Neural Network mappings
- Live Concatenative Synthesis
- Path find through multidimensional space of laptop improvisation interface (track MIDI & OSC data, MIR data)
- tedbot



Thank you. Questions?

