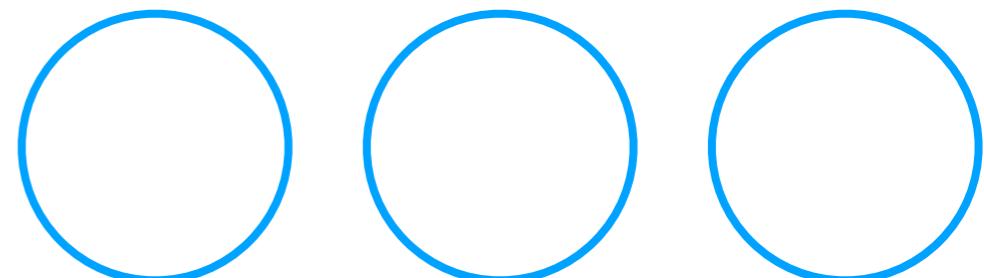


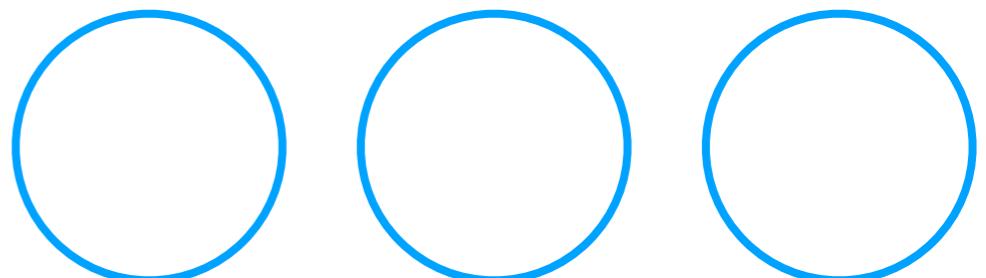
Approaches to Live Performance and Composition with Machine Learning and Music Information Retrieval Analysis

Ted Moore
CHIMEFest 2019



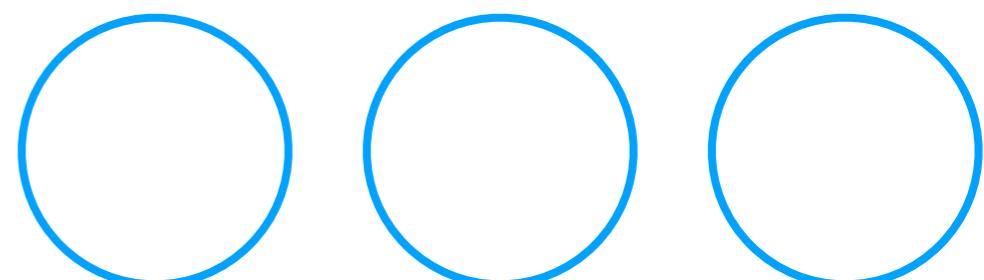
My Practice

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



Interest in Music Information Retrieval & Machine Learning

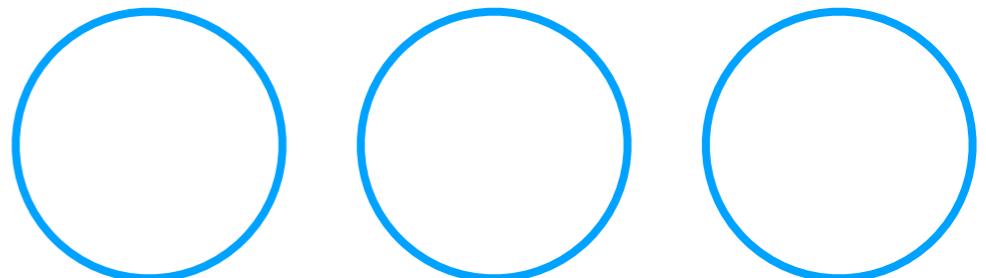
- MIR workshop at CCRMA summer 2018
- In what new ways can I approach sound?
- What can an algorithm do for (with) me? What can it tell me?
- Computational thinking
- What other routes are there to the same goal?



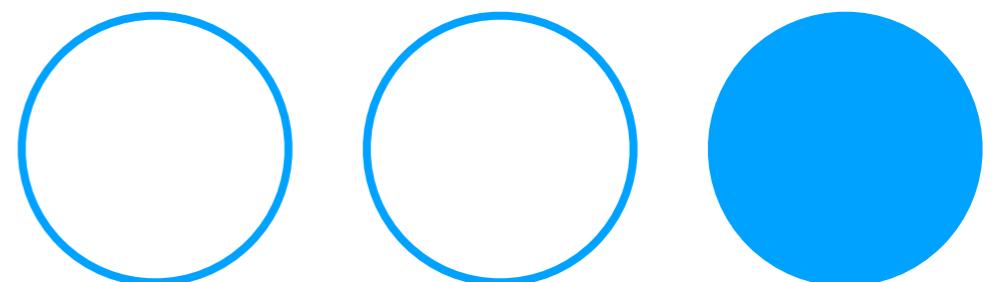
What is the goal?

Make sounds and forms that I find artistically compelling.

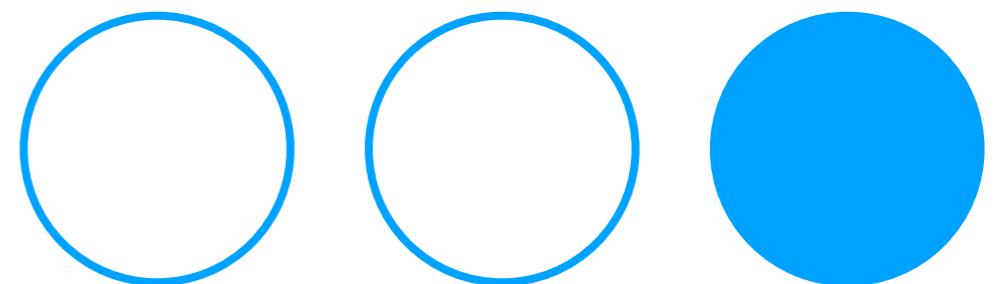
Today I share 3 examples of using these tools in that pursuit.



1. Gestural Control in MIR Space



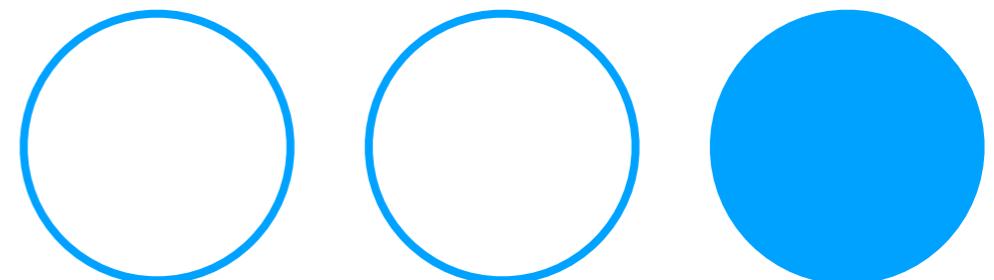
playing random grains from a
collection of samples



Music Information Retrieval Class in SuperCollider

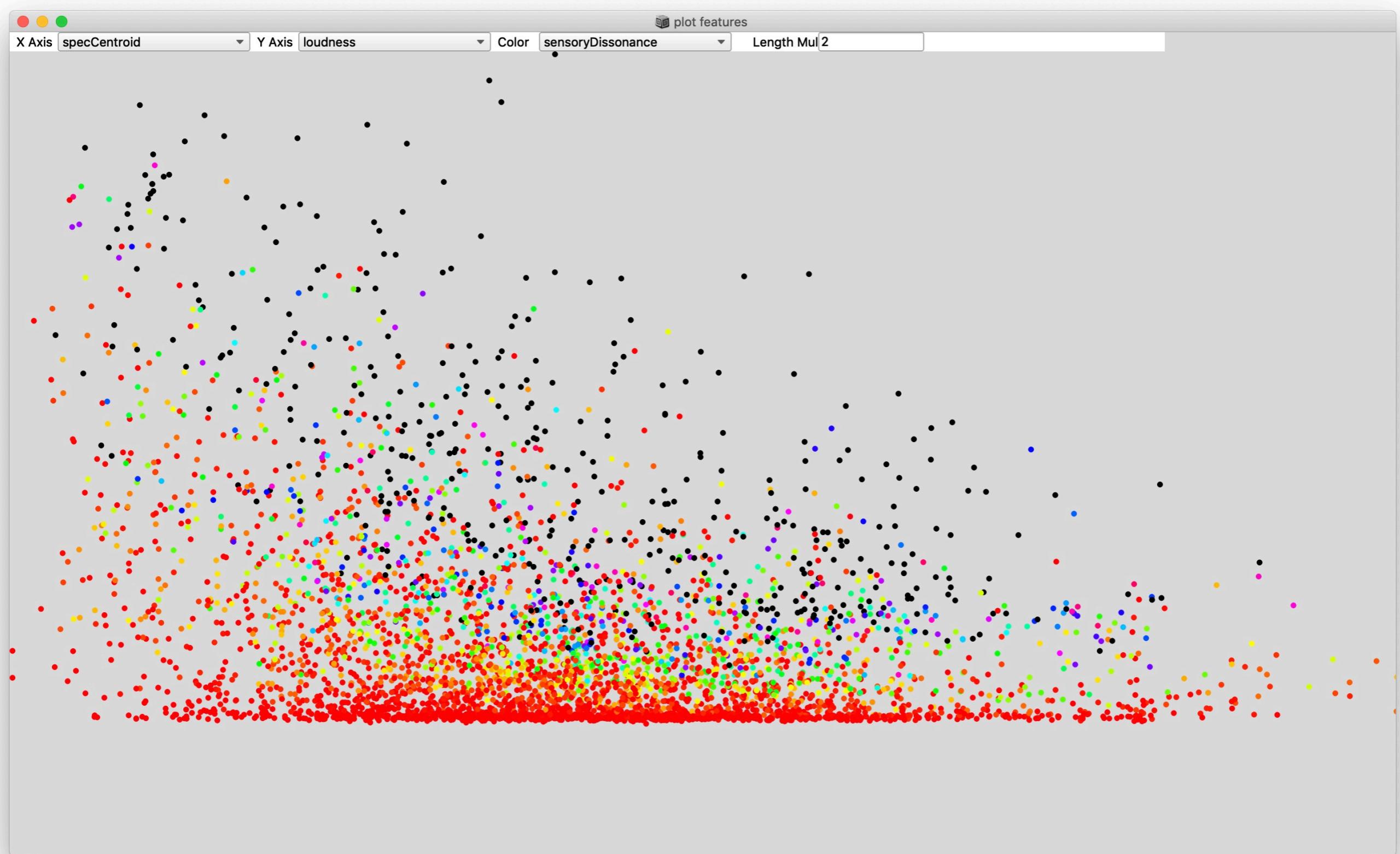
- 23 Dimensions of Analysis
- Onset Detection
- NRT Analysis of Files and Corpus
- Live Analysis along same Dimensions
- Returns MIRAnalysisFile, an object of its own

0: amplitude	13: mfcc04
1: fftCrest	14: mfcc05
2: fftSlope	15: mfcc06
3: fftSpread	16: mfcc07
4: loudness	17: mfcc08
5: sensoryDissonance	18: mfcc09
6: specCentroid	19: mfcc10
7: specFlatness	20: mfcc11
8: specPcile	21: mfcc12
9: zeroCrossing	22: mfcc13
10: mfcc01	
11: mfcc02	
12: mfcc03	



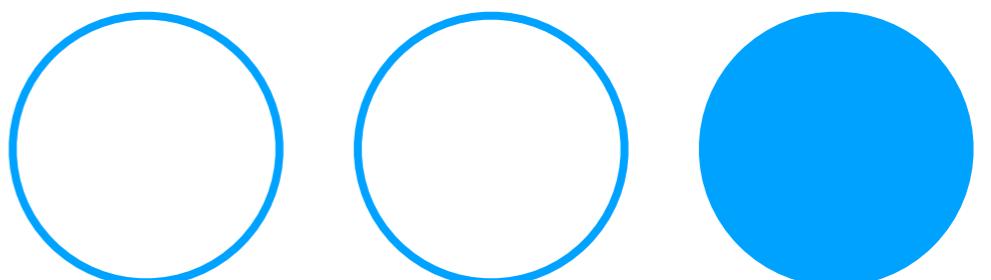
MIRCorpus class in SuperCollider

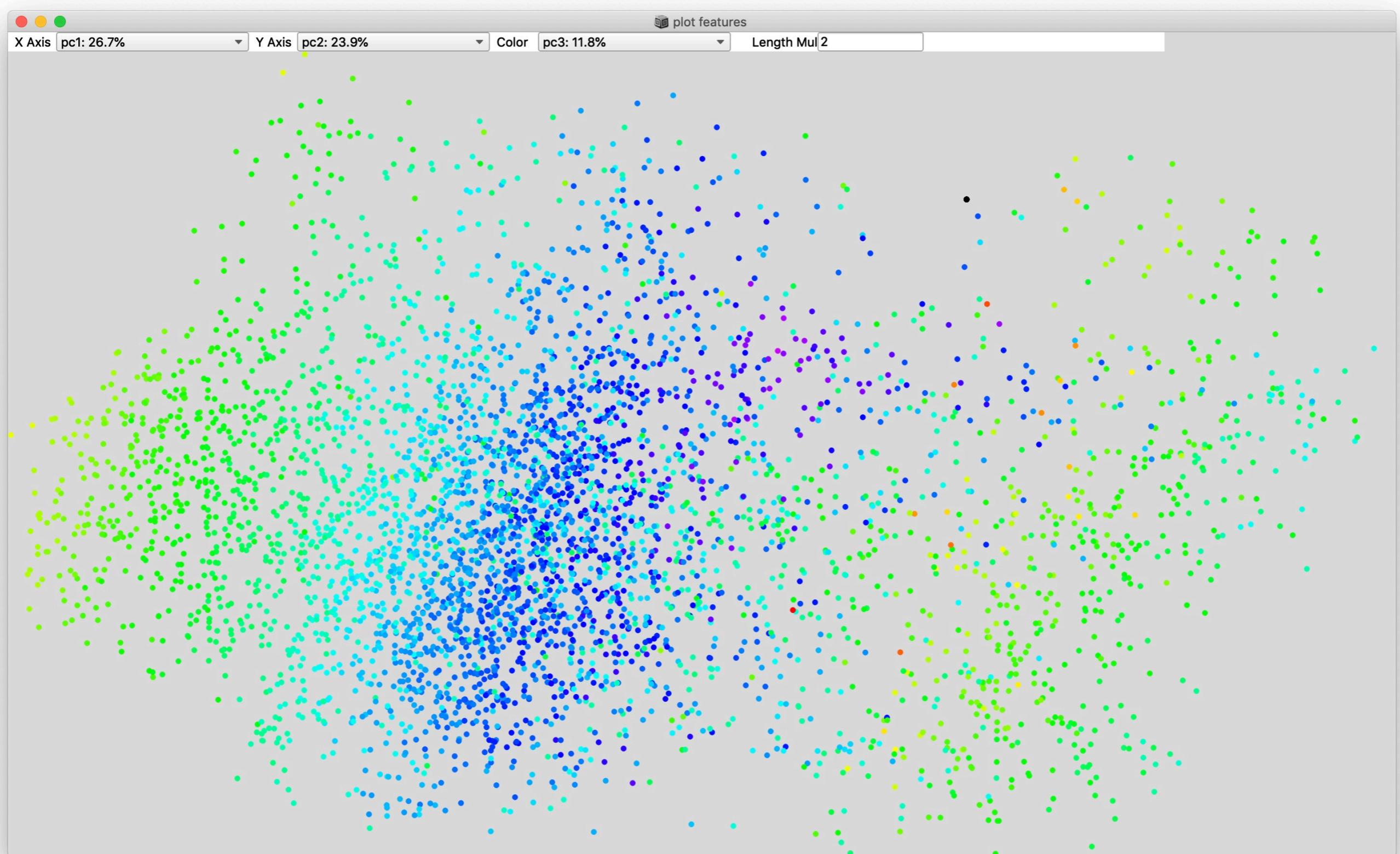
- MIRAnalysis on source corpus
- NRT
- Returns MIRCorpusItem



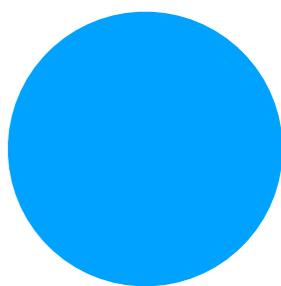
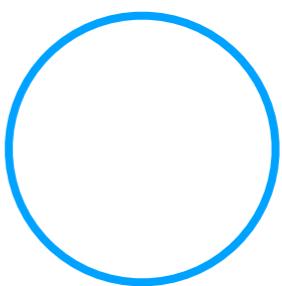
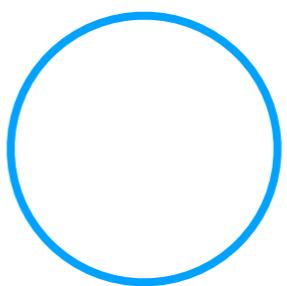
Principal Component Analysis

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

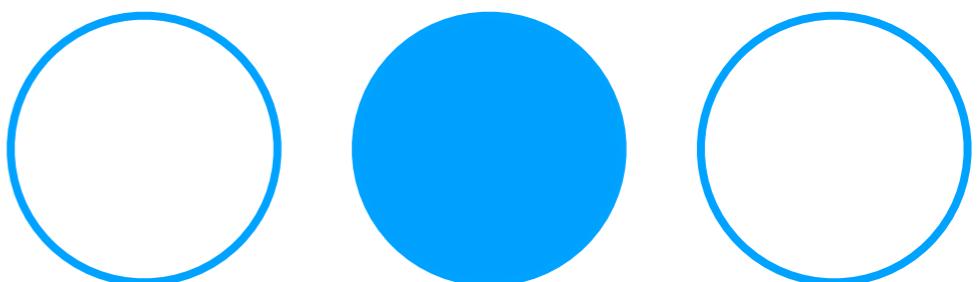




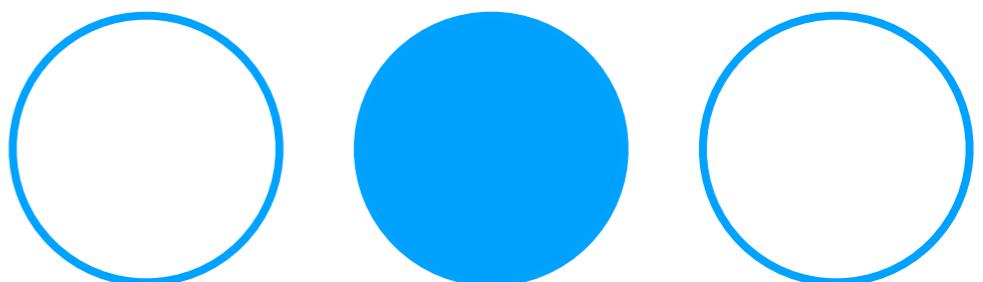
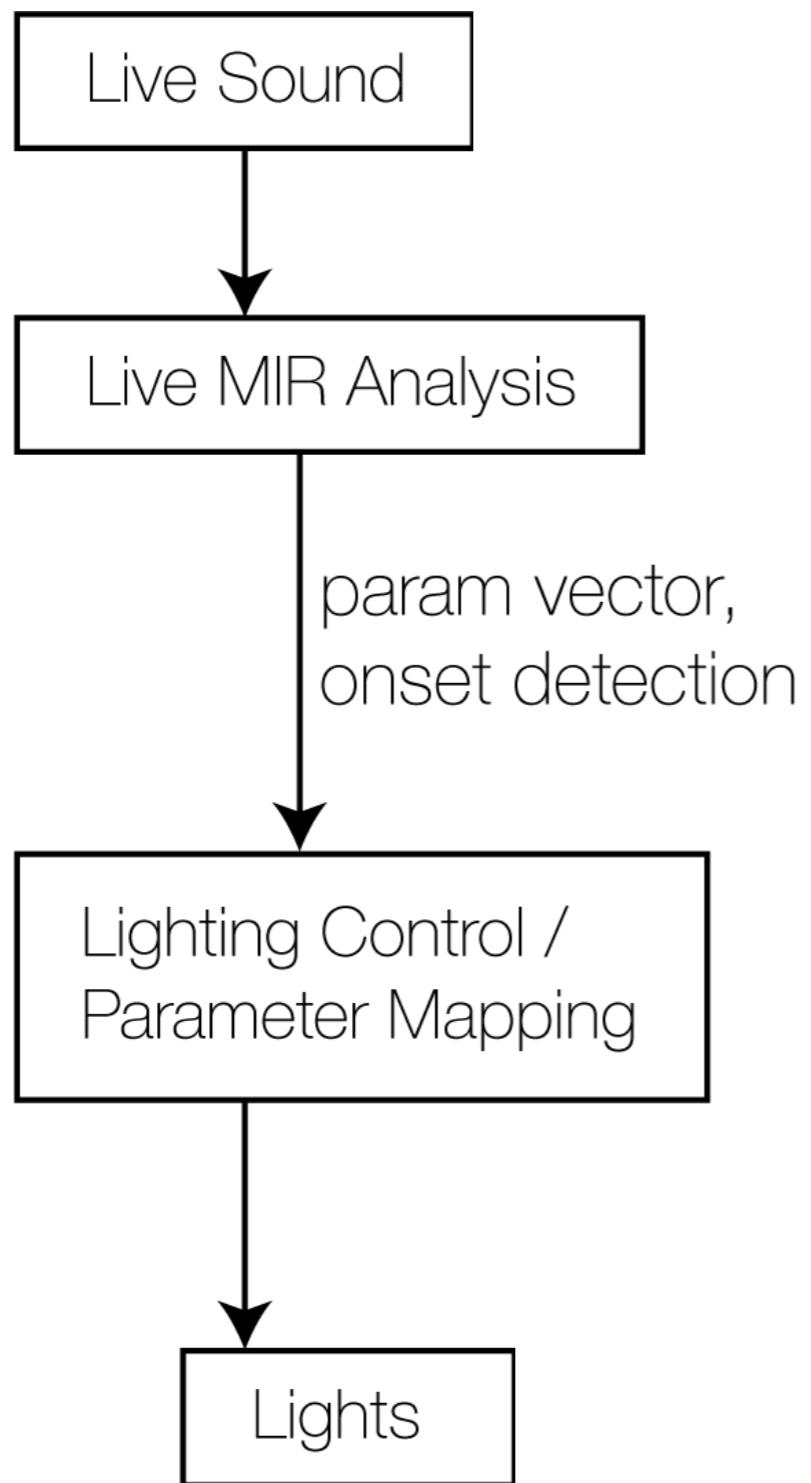
demo time



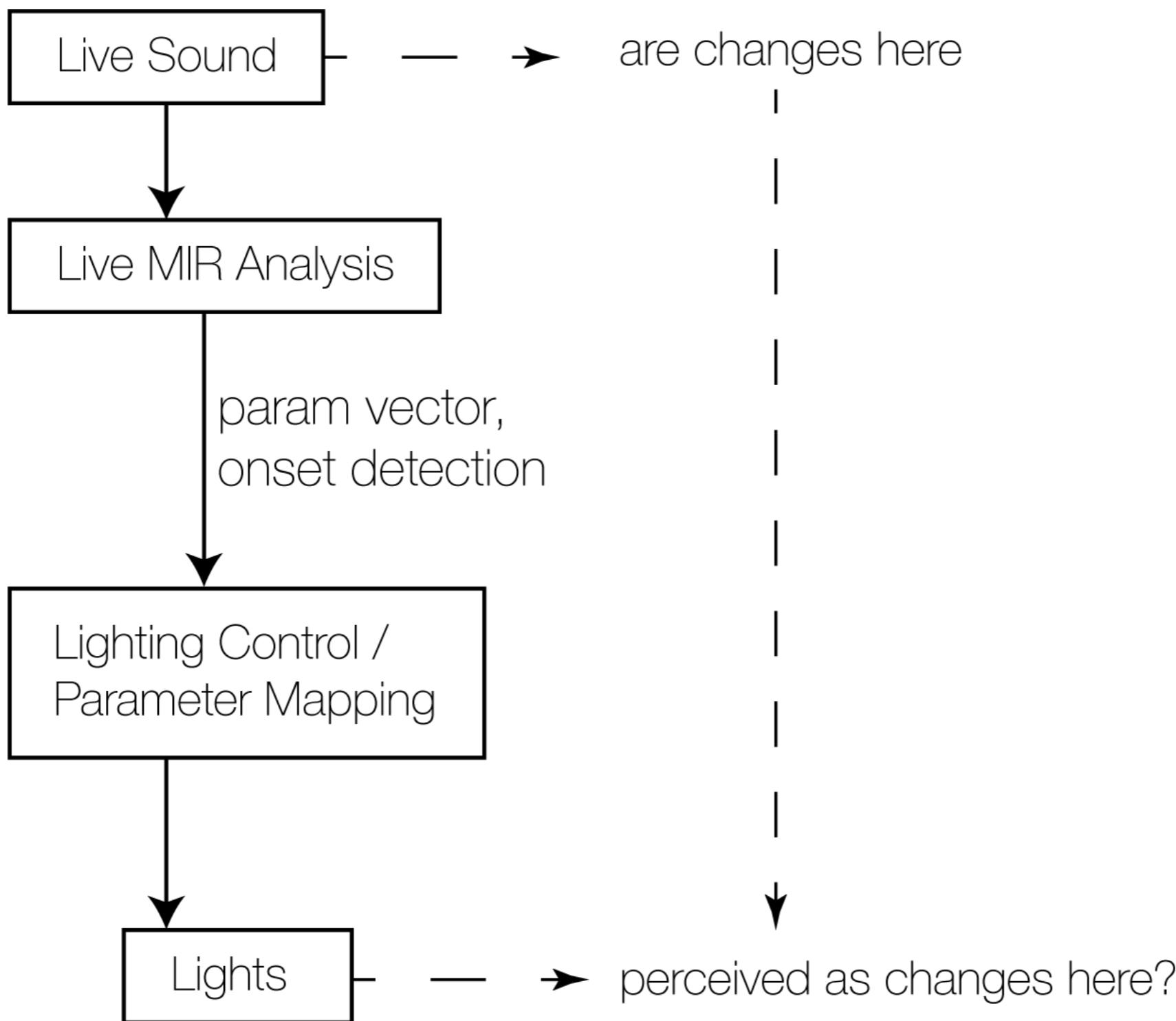
2. Live Sound Classification



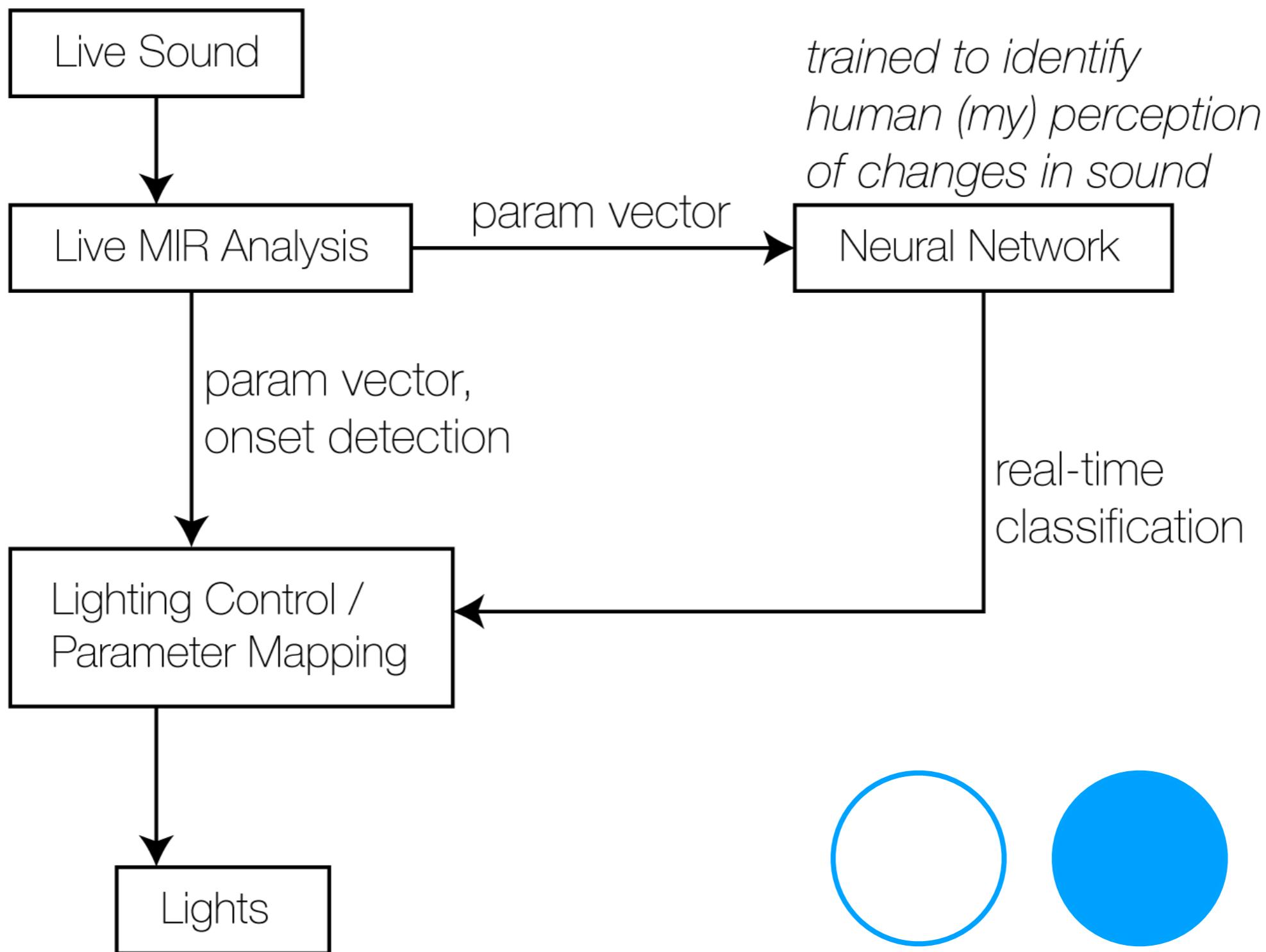
Machine Listening System



Machine Listening System



Machine Learning System





distorted_noise



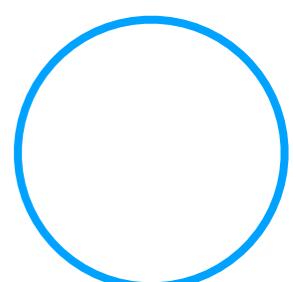
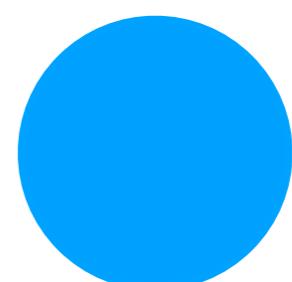
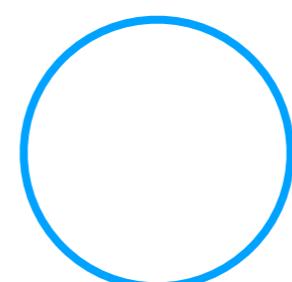
high_squeal



low_impulses



sus_noise_quiet

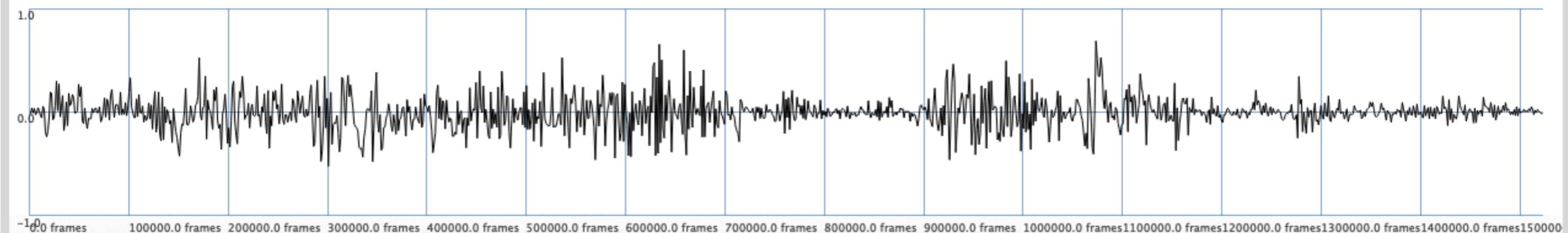
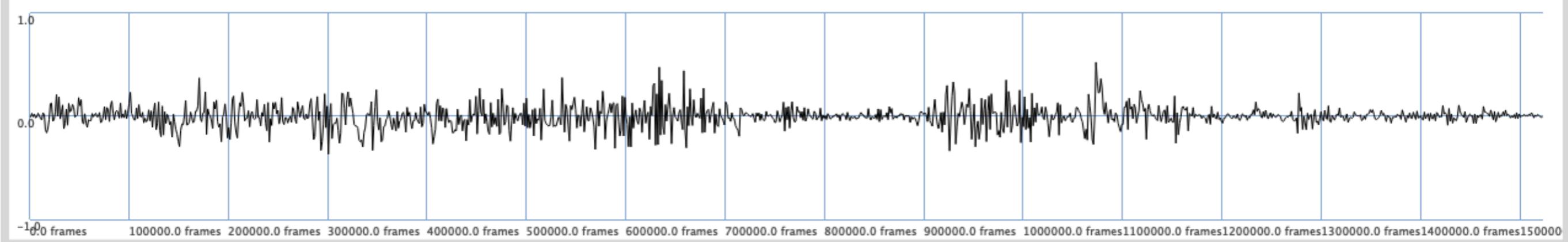


```
1 NeuralNetwork {
2     var <>net, <>learningRate, e = 2.71828, <shape, <>activation, <>normalizedRanges;
3
4     *new {
5         arg shape,learningRate = 0.05,activation = "relu",normalizedRanges;
6         ^super.new.init(shape,learningRate,activation,normalizedRanges);
7     }
8
9     init {
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        });
30    }
31}
```

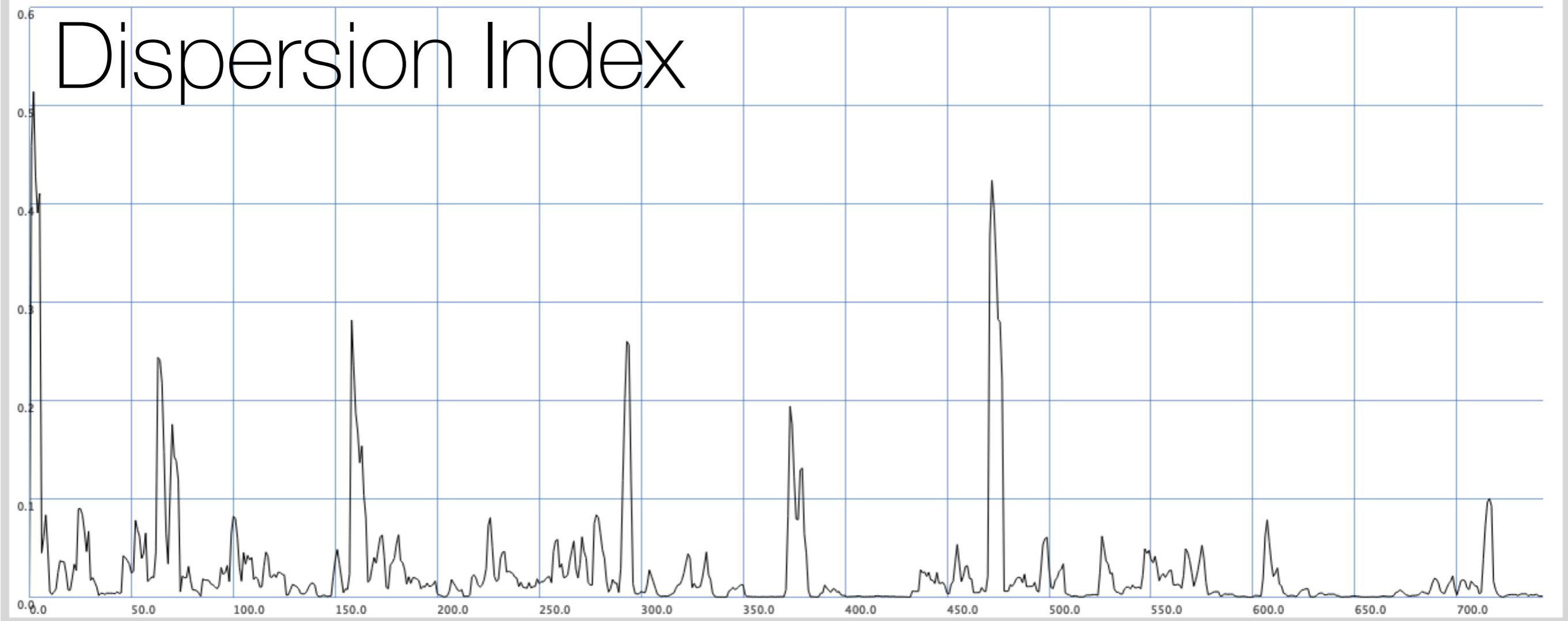


```
feedLights = FeedLightMaster([

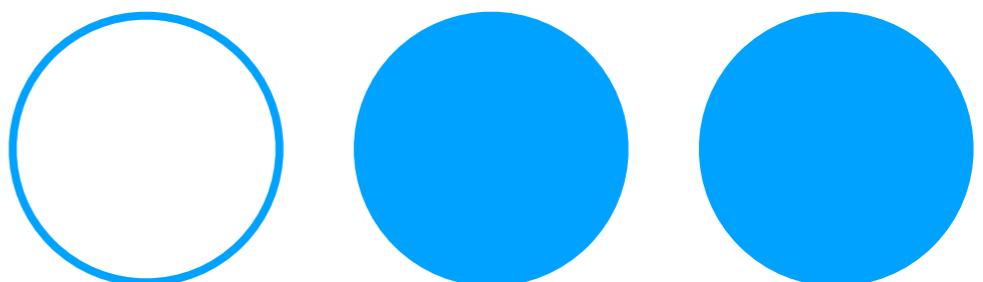
    // distorted noise
    FeedLightMode(nLights,[
        FeedLightGroup([
            \amplitude,\myAmp,\v,ControlSpec(0.01,1,\exp),
            \specCentroid,ControlSpec(50,5000,\exp),\h,ControlSpec(0.5,0.7),
            \specFlatness,nil.asSpec,\s,ControlSpec(1,0.3)
        ]),
        FeedLightGroup([
            \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



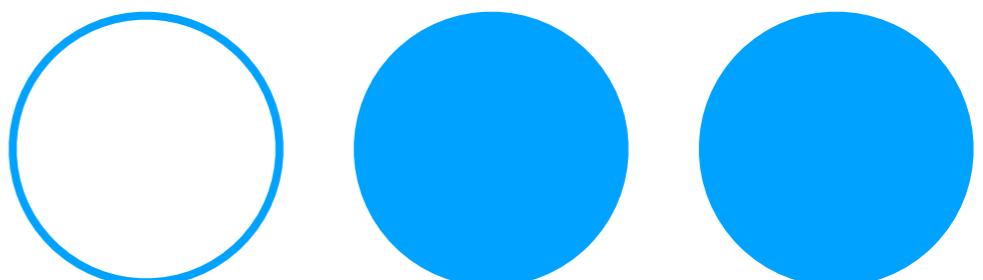
3. Corpus Concatenation



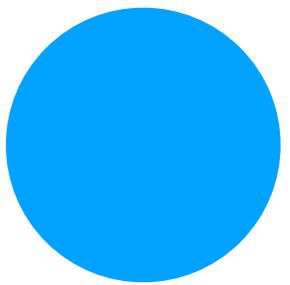
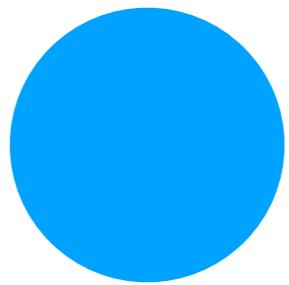
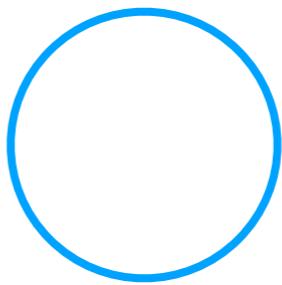
ConcatSynthNRT class in SuperCollider

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

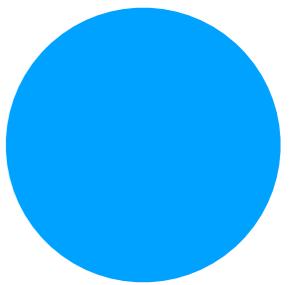
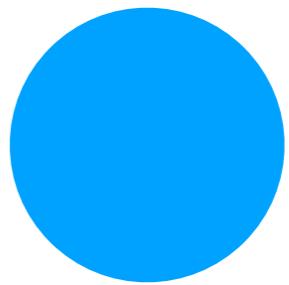
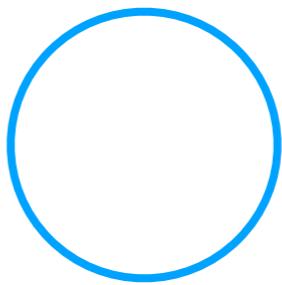
- renderFromCorpusAndFilePath()
 - kNN
- renderFromCorpusAndRawFrames()
 - kNN
- renderFromArrayOfCorpusItems()
 - from given path



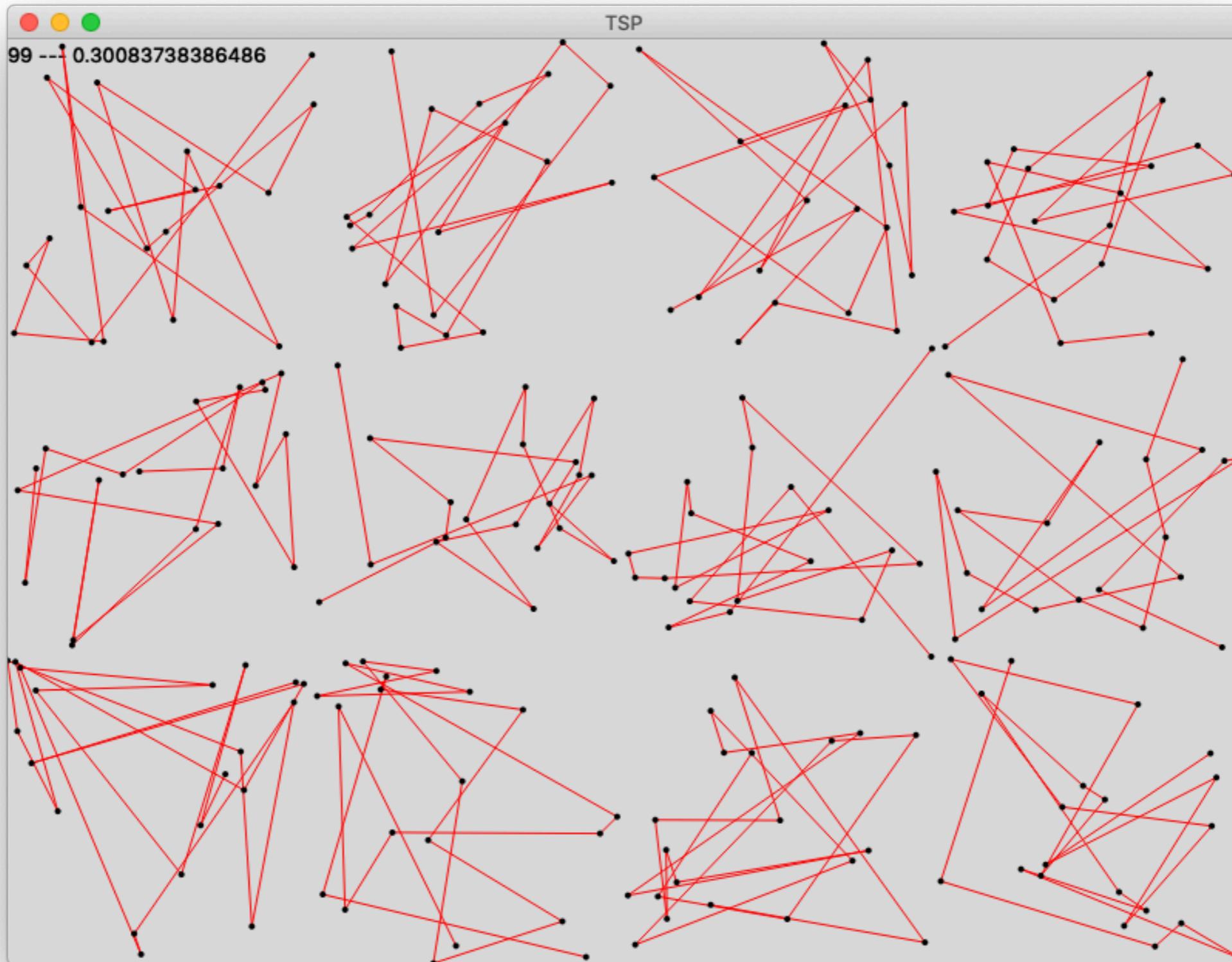
what does the fox say?



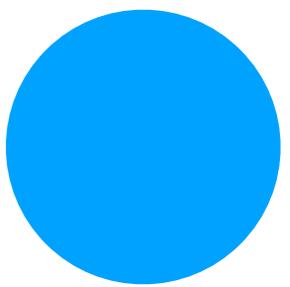
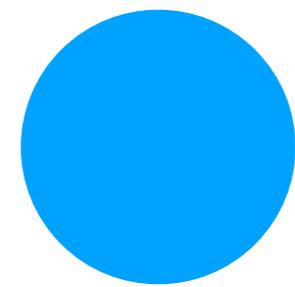
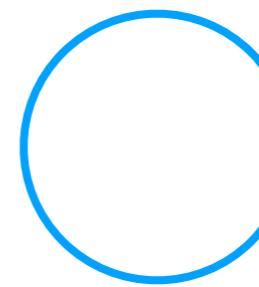
what does the fox say?

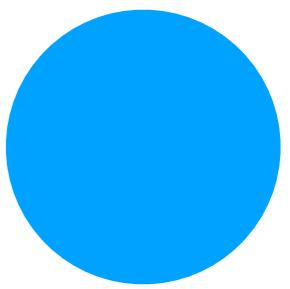
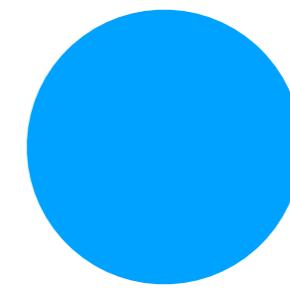
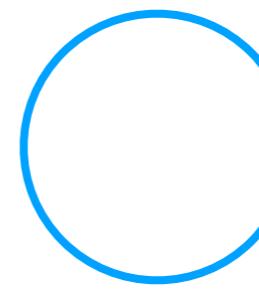
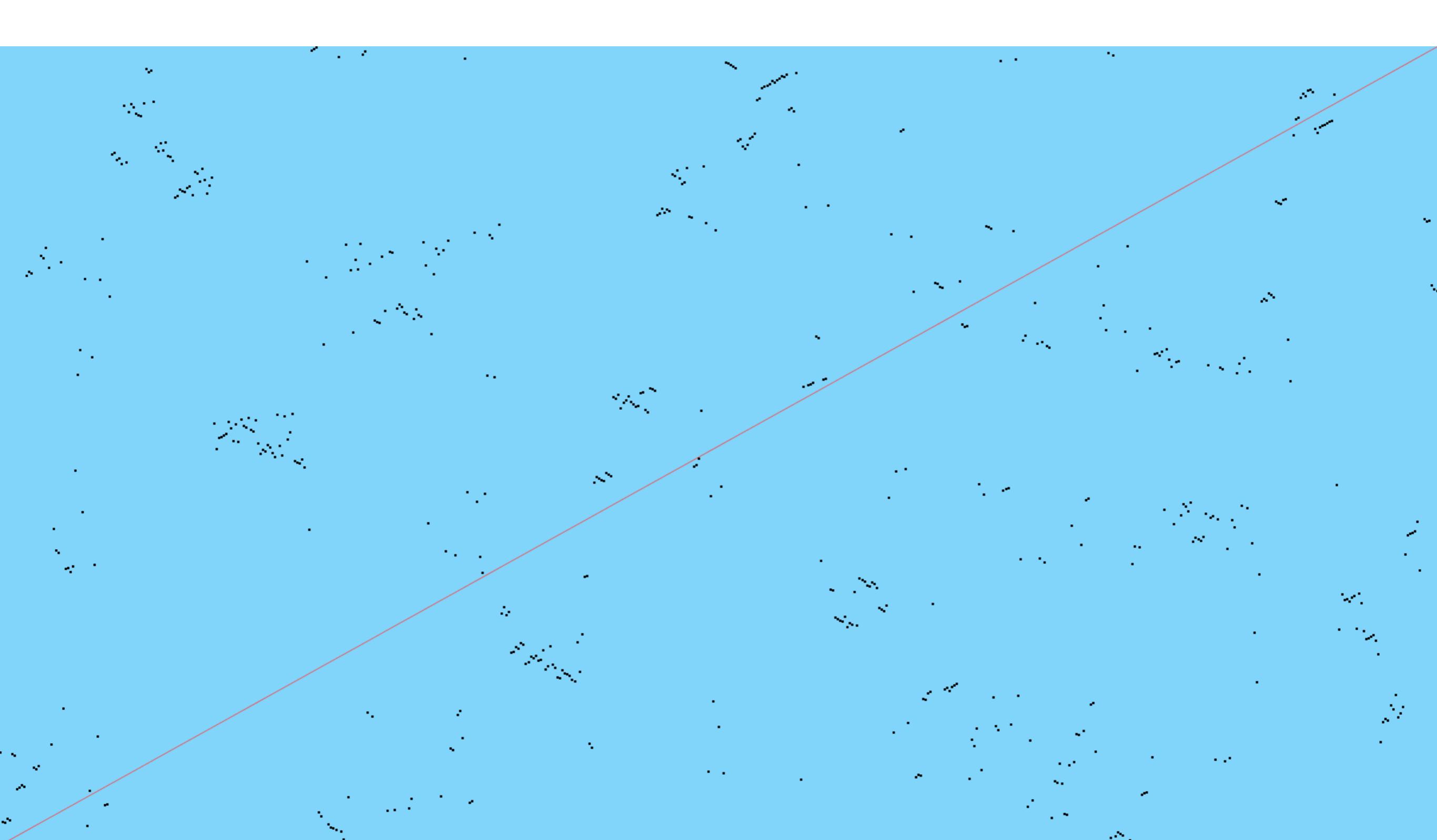


pathfinding through MIR space: empowering algorithms to organize time

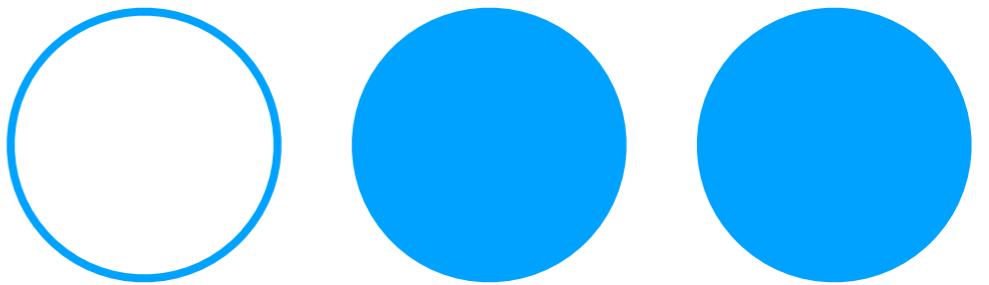


what does the sax say?





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

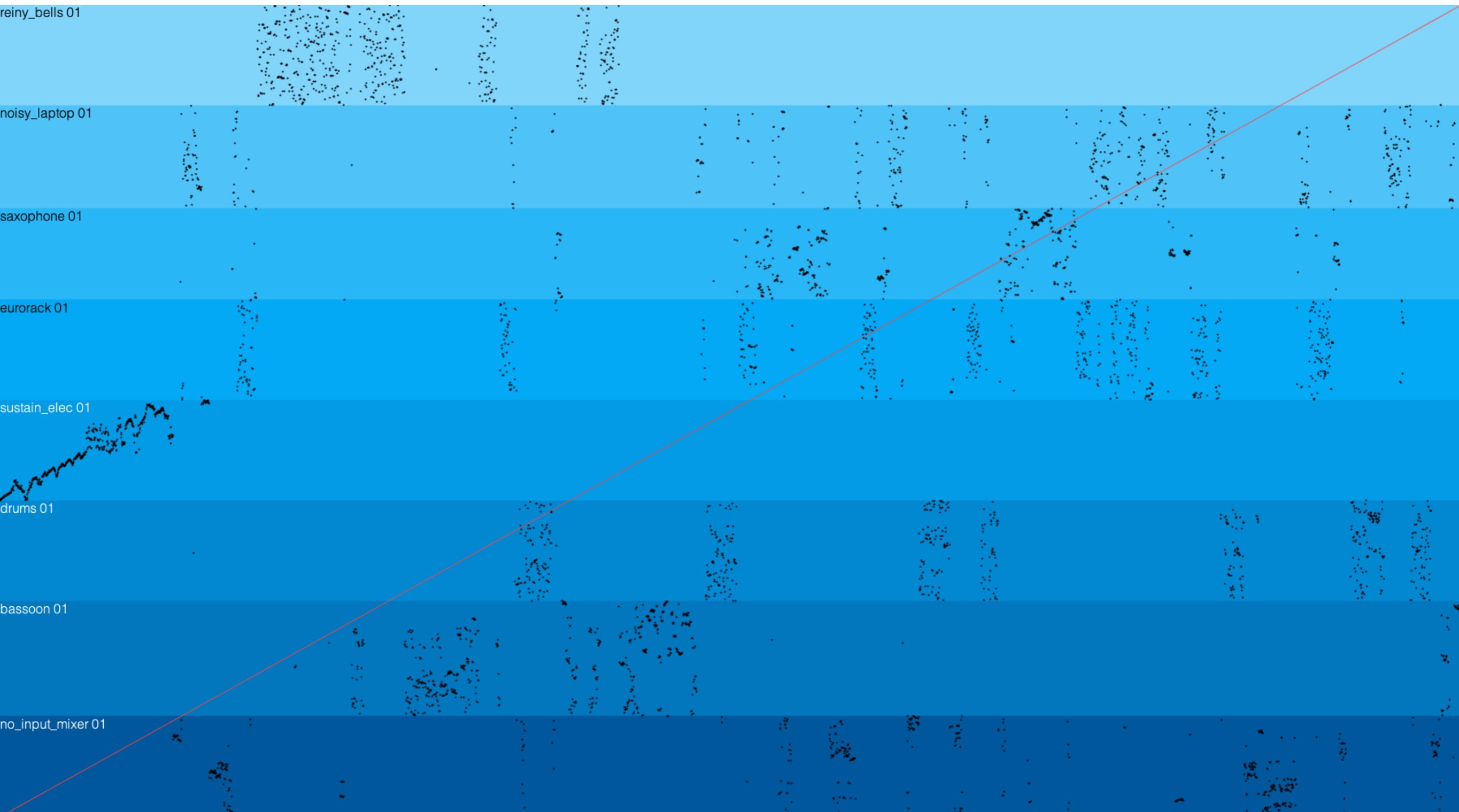
sustain_elec 01

drums 01

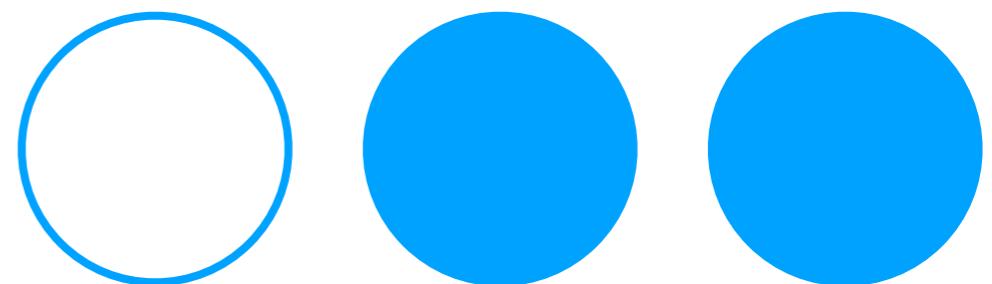
bassoon 01

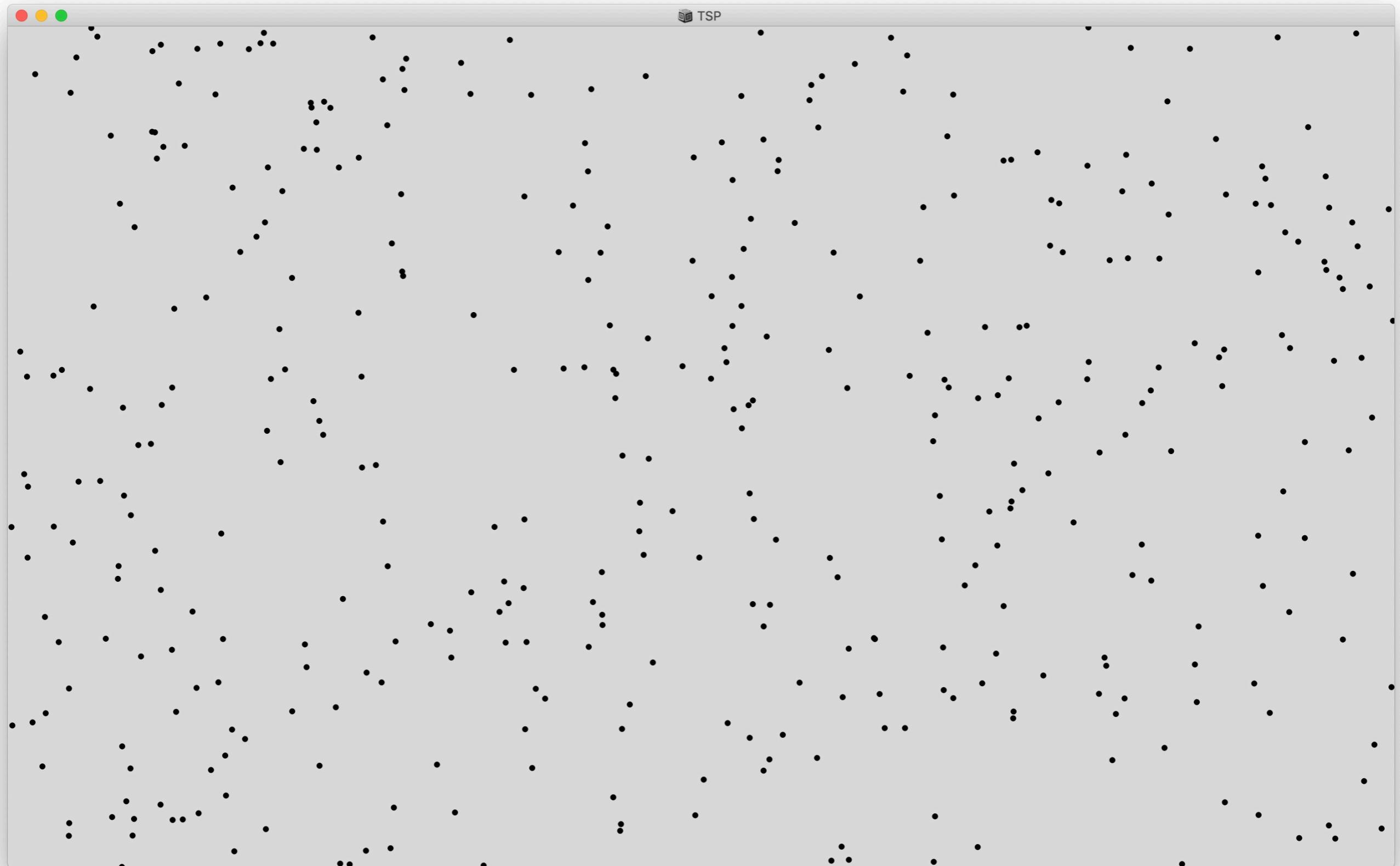
no_input_mixer 01

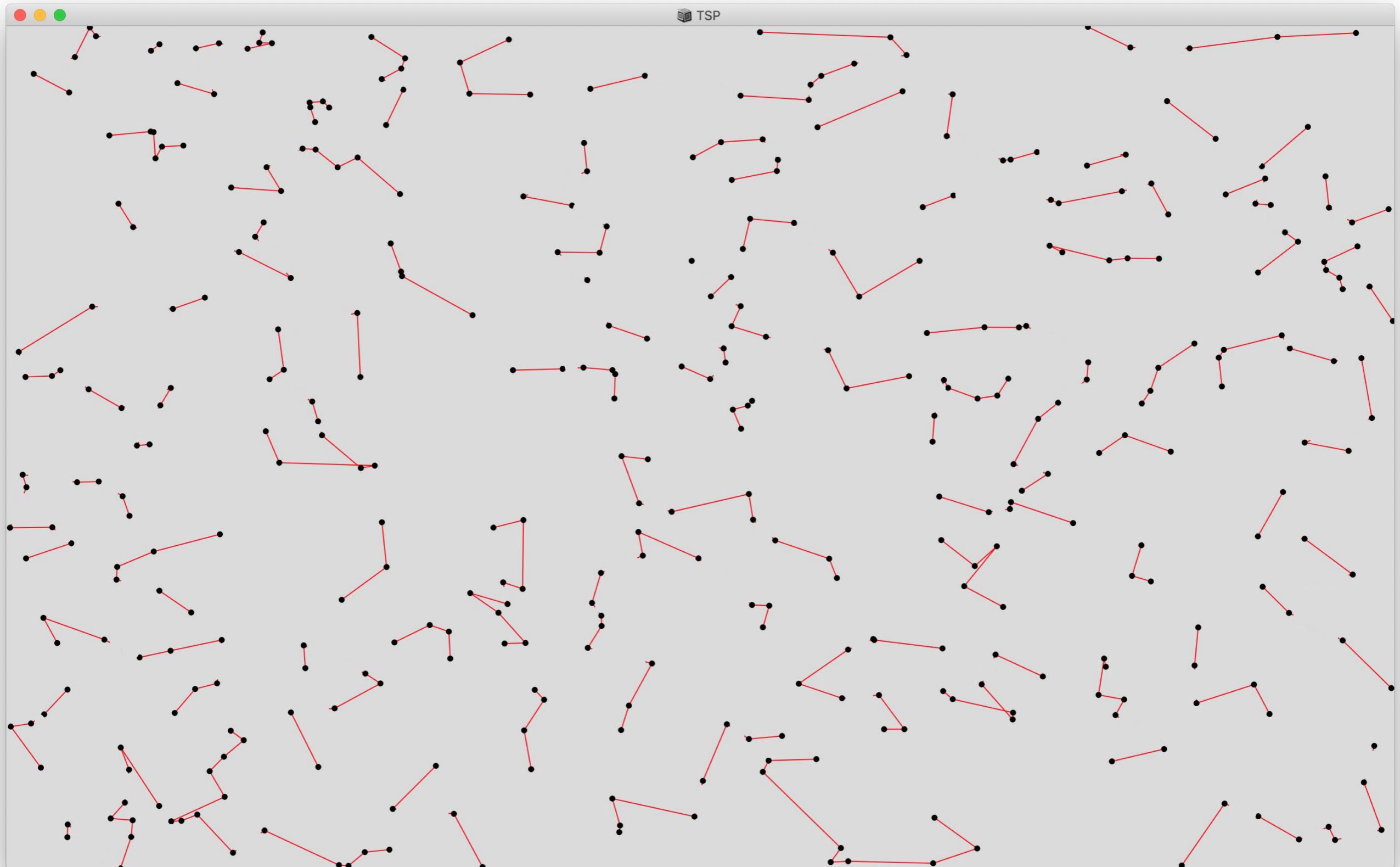


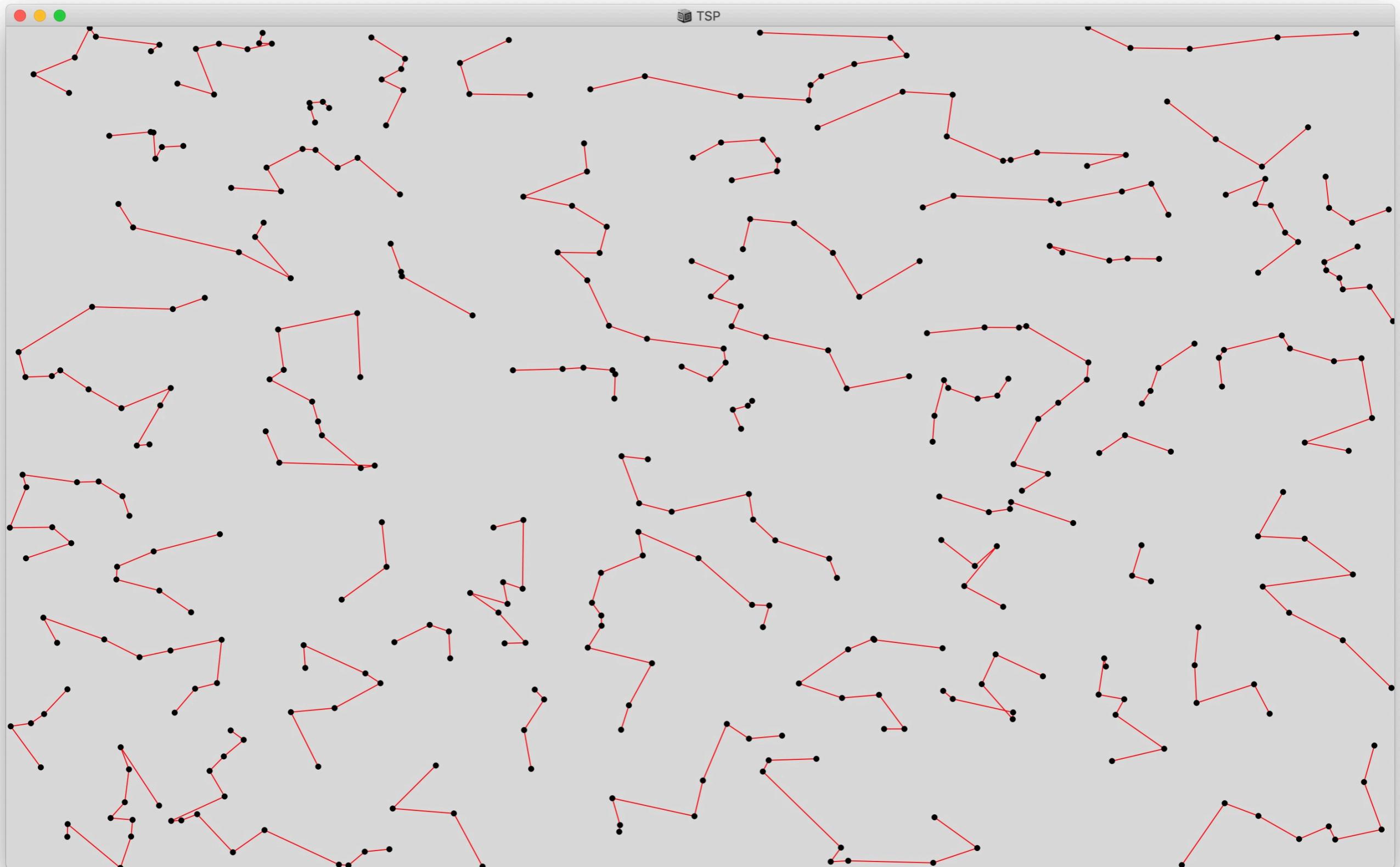


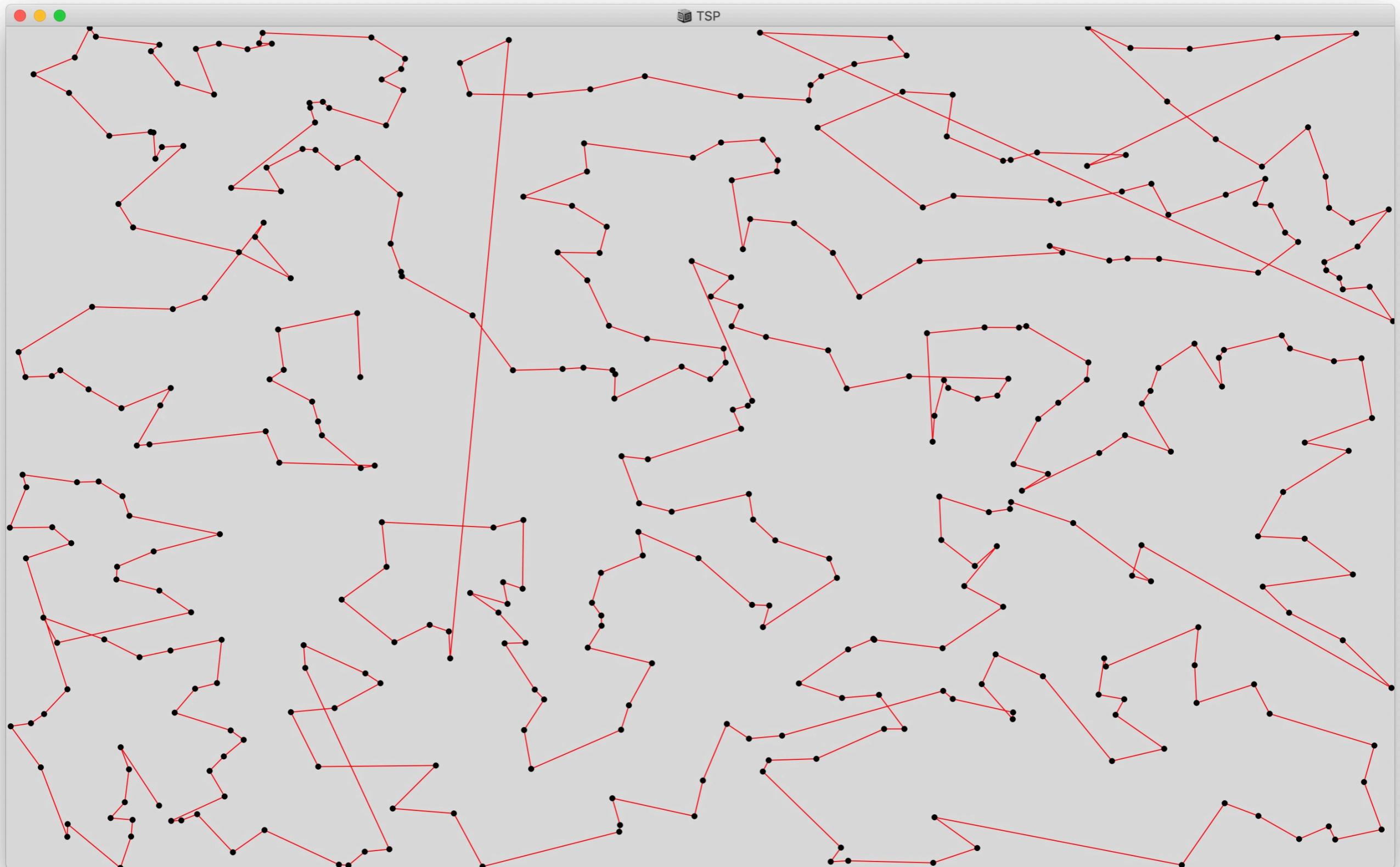
machine composed timbral
fusion, gesture, & form











pathfinding through FFT space: towards machine learning “synthesis”

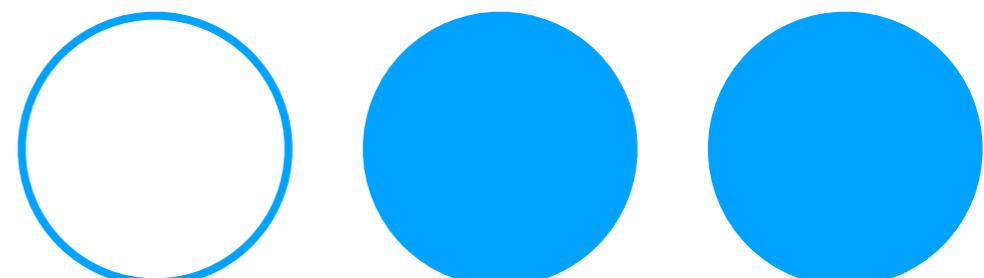
```
1 FFTNRT {  
2  
3     *fft {  
4         arg filePath, action, fftSize = 2048, overlap = 2;  
5         SoundFile.use(filePath,{  
6             arg sf;  
7             var data, window, frames, currentSample = 0, hopSamples, fft, imag;  
8  
9             hopSamples = fftSize / overlap;  
10  
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”

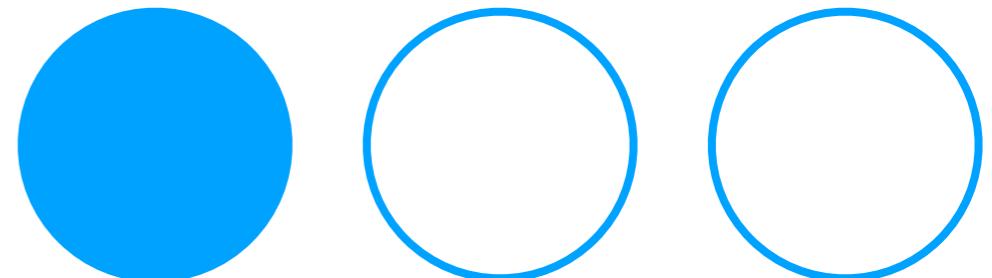
```
Python running for 08:53:33.471
Python running for 08:53:34.471
Python running for 08:53:35.471
[ 6982, 6981, 4968, 8067, 8002, 8001]
```

saxophone, no input mixer, laptop, eurorack



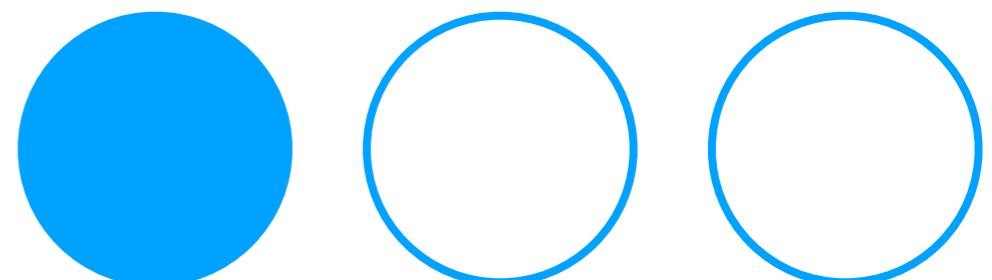
“The view according to which the novelty of a work guarantees its quality is often expressed in electroacoustic music circles, and for some it is the only criterion of worthiness.”

–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?

