

# Learning music production practice through evolutionary algorithms

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# Intelligent Music Production

- The field of intelligent music production has been an active research topic for over a decade.
- The aim is to develop systems which are capable of performing common tasks in music production, such as level-balancing, equalisation, panning, dynamic range compression and application of artificial reverberation.

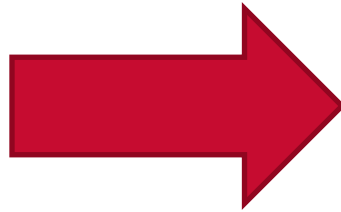
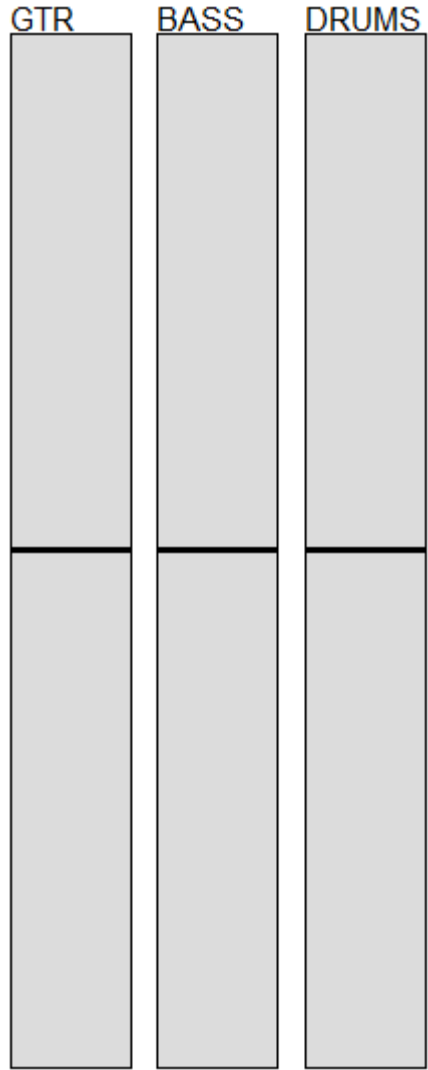
# Intelligent Music Production

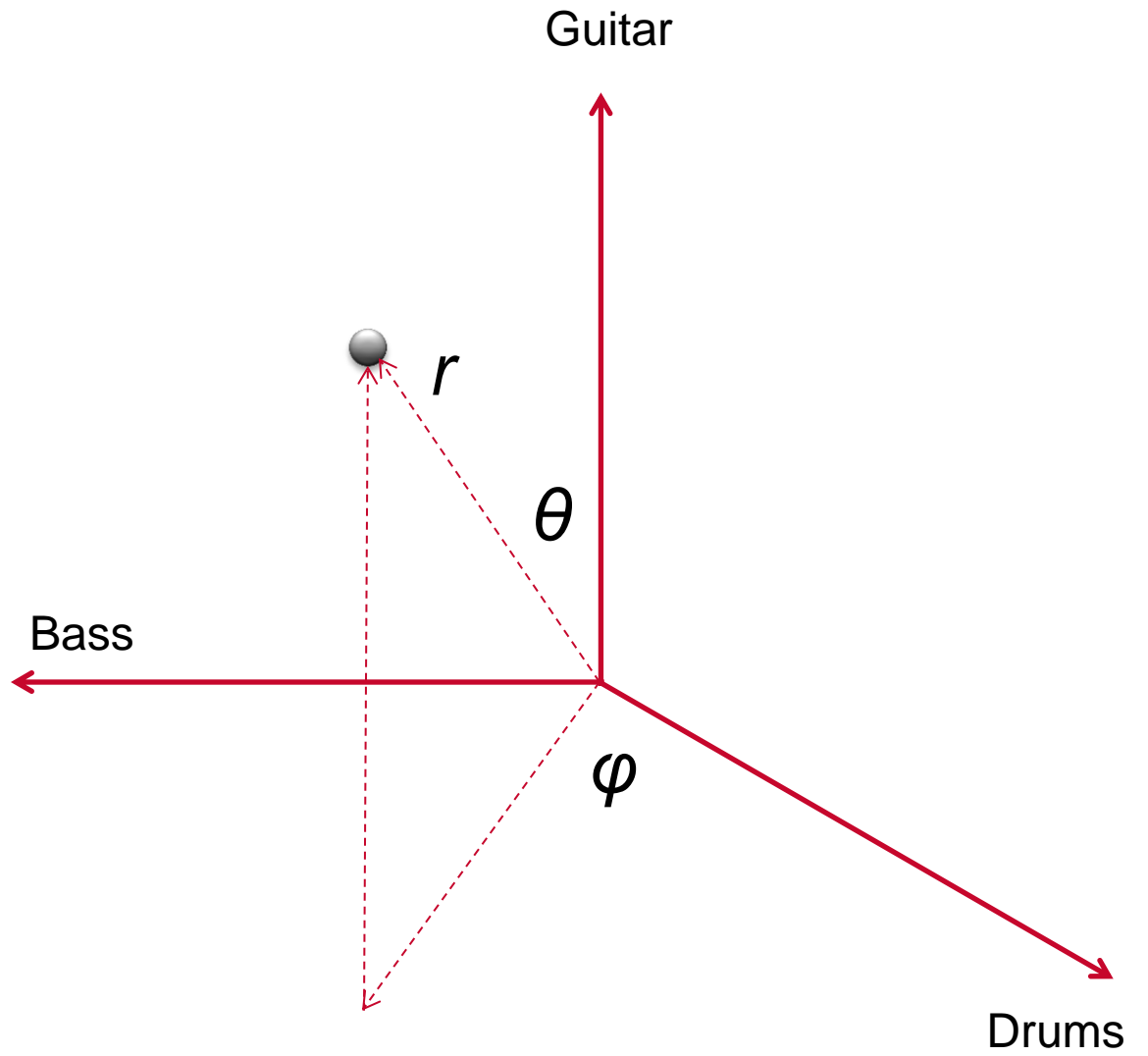
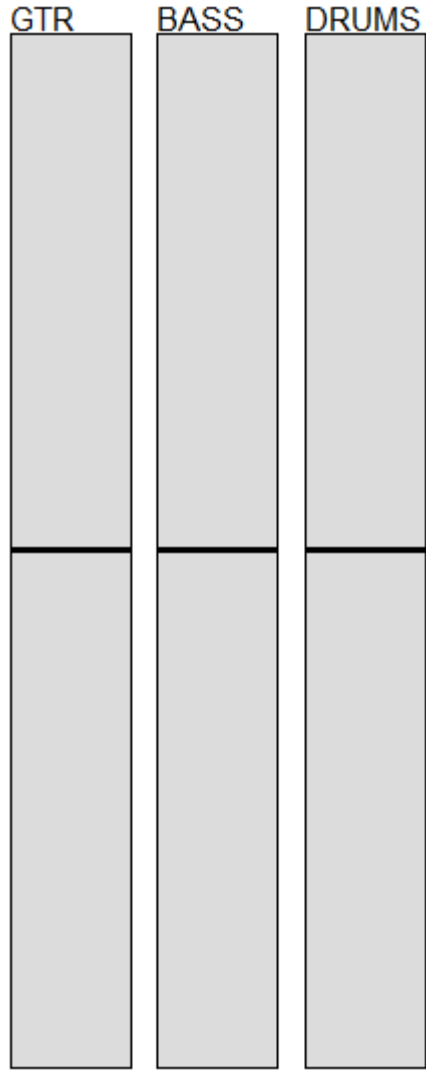
## Typical system aesthetics and constraints

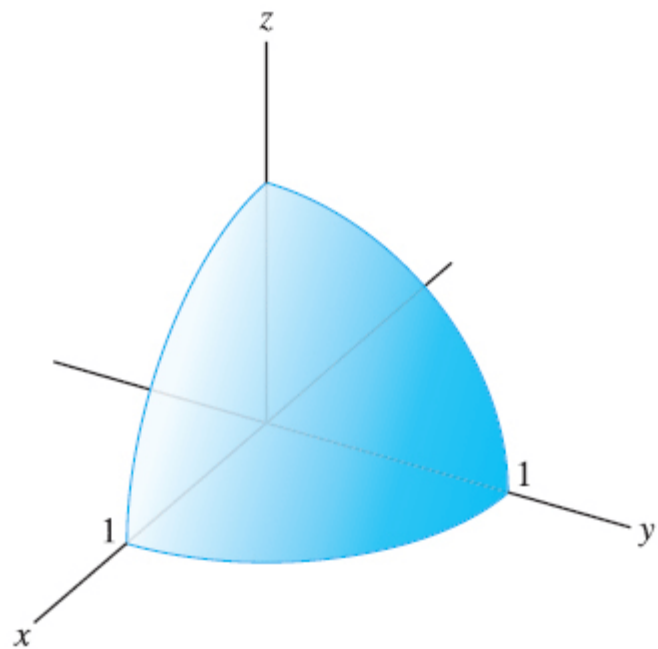
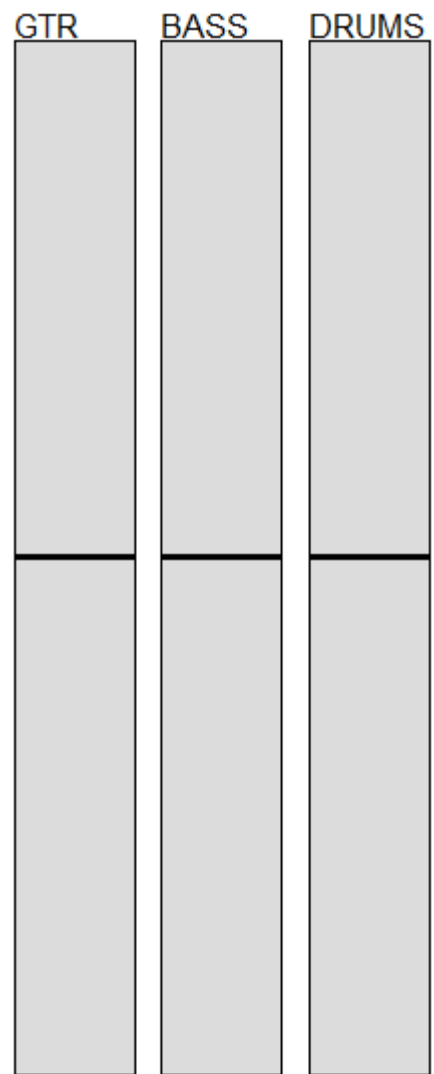
- Real-time (designed for “live” environment)
- Applies audio-specific processing, based on “best-practice”
- Searches for the “best” mix, **objectively**, irrespective of user

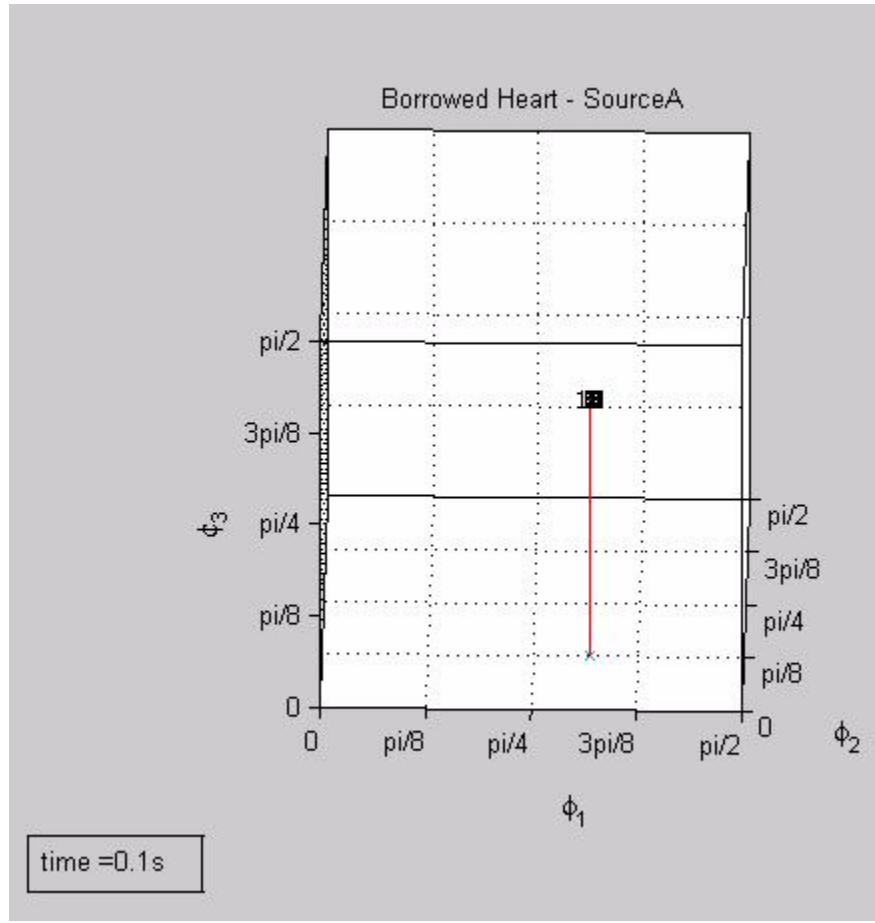
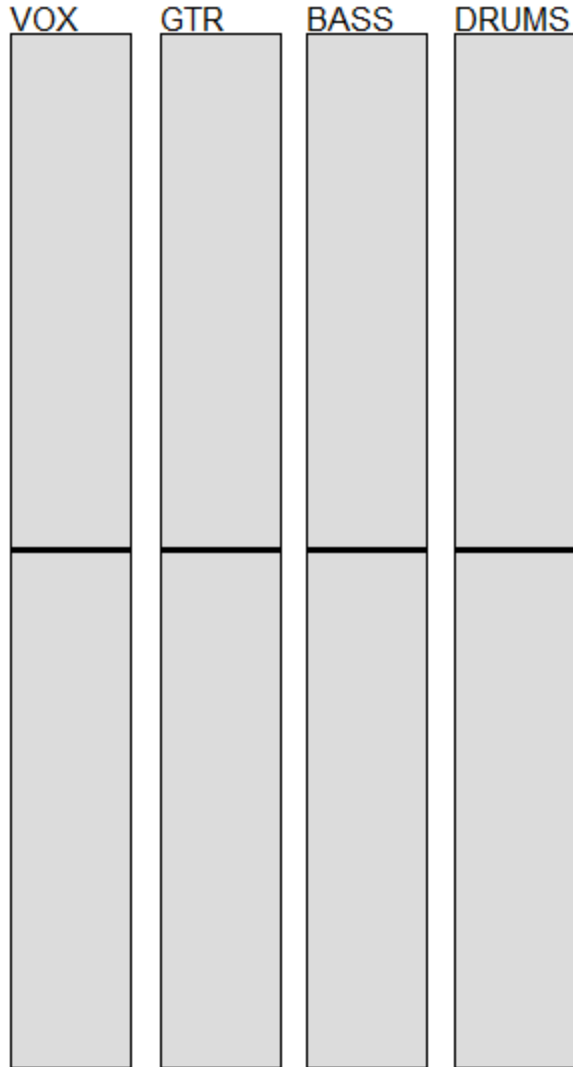
## Proposed system aesthetics and constraints

- Off-line (designed for “studio” environment)
- Makes no assumptions about “best-practice”
- Searches for the “best” mix, **subjectively**, according to the user

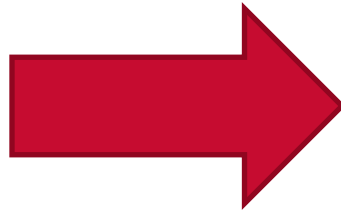
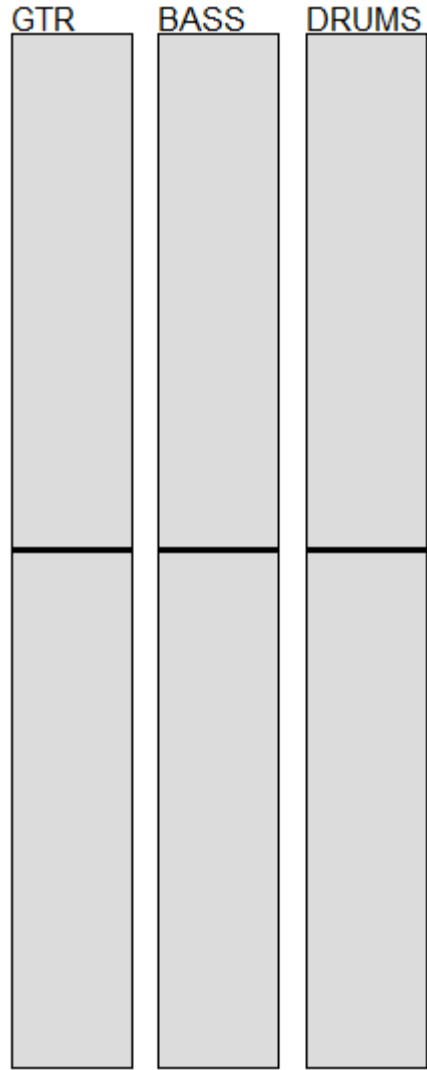




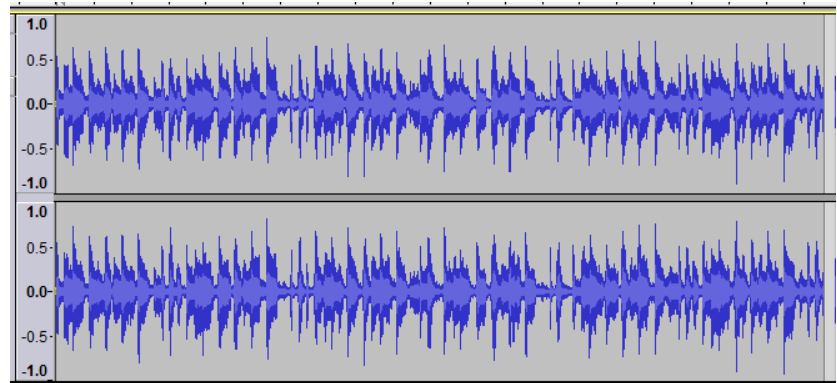
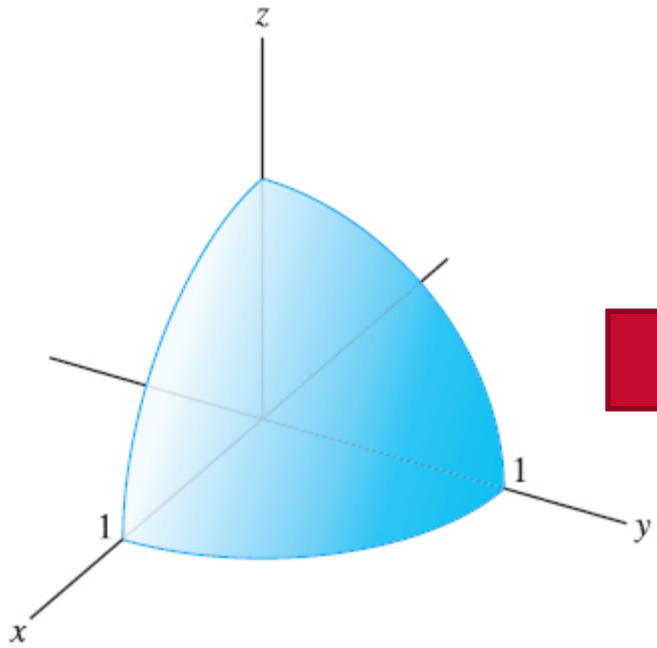




“Navigating the mix-space: theoretical and practical level-balancing technique in multitrack music mixtures”, **Wilson/Fazenda, 2015**

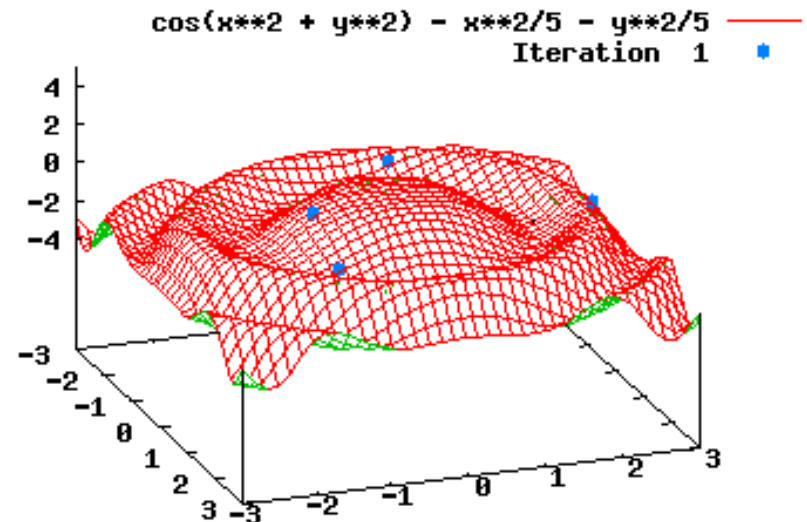






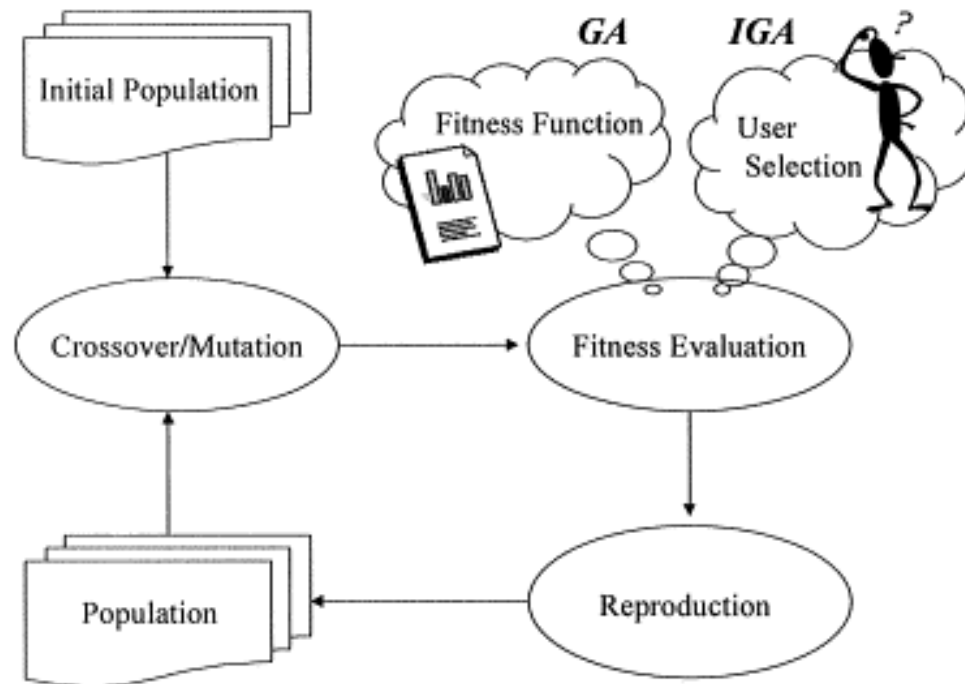
# Evolutionary Algorithms

- Generic population-based metaheuristic optimisation algorithm
- Can be used for global optimisation
- Techniques include:
  - **Genetic algorithm**
  - Particle Swarm Optimisation
  - Ant Colony Optimisation
  - Bees Algorithm



# (Interactive) Evolutionary Algorithms

- Particularly suitable to aesthetic design problems
  - Non-linear, not deterministic

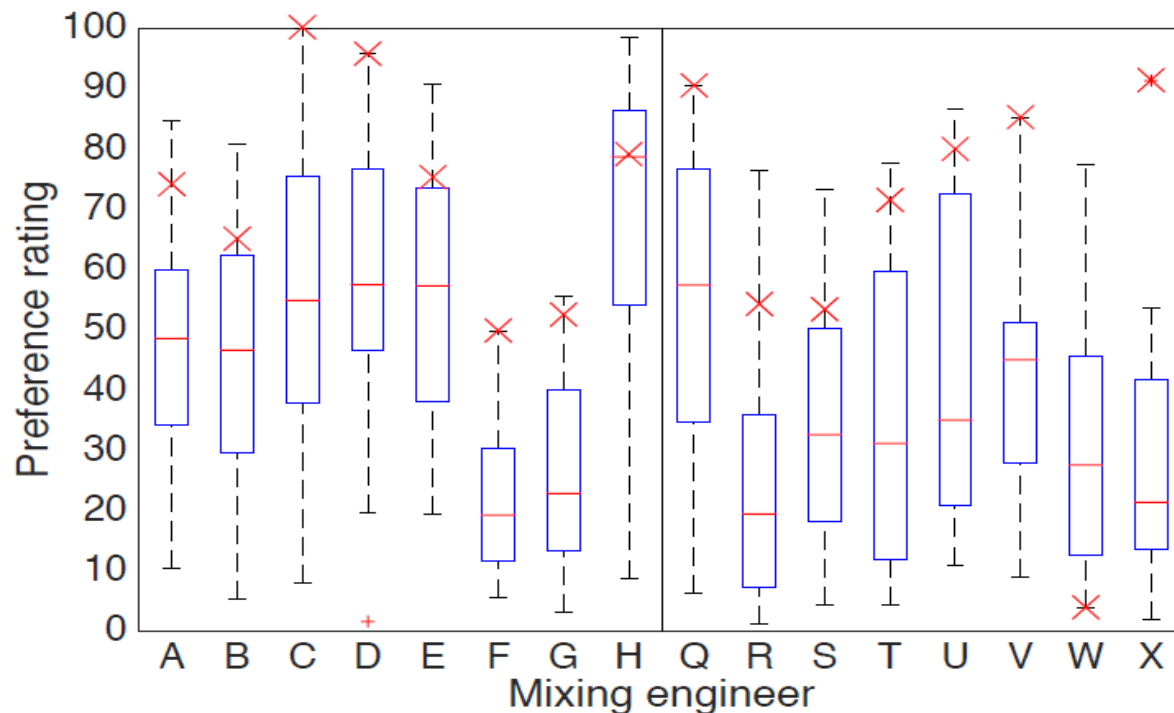


# (Interactive) Evolutionary Algorithms

- Particularly suitable to aesthetic design problems
  - Non-linear, not deterministic
- IEA has applied to...
  - Fashion design (Kim/Cho, 2000)
  - Structural design (O'Neill et al. 2010)
  - Logo design (O'Neill/Brabazon, 2008)

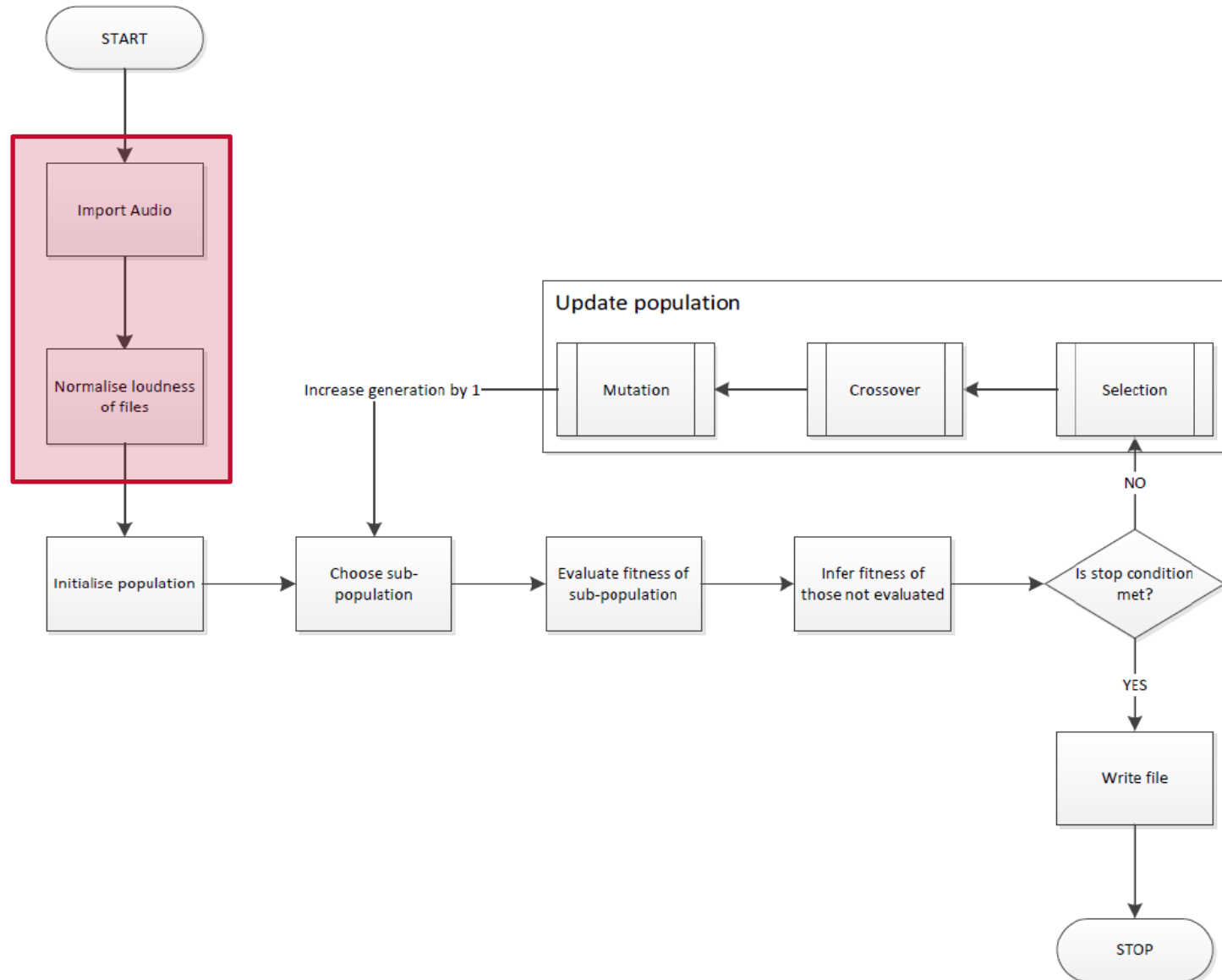
# (Interactive) Evolutionary Algorithms

- Why have user-assisted system?
  - The “best” mix won’t be the best for everyone
  - The user can help guide the mixing system towards their ideal mix



*“Perceptual Evaluation of Music Mixing Practices”,  
De Man et al., 2015*

# Flowchart



# Import and normalise

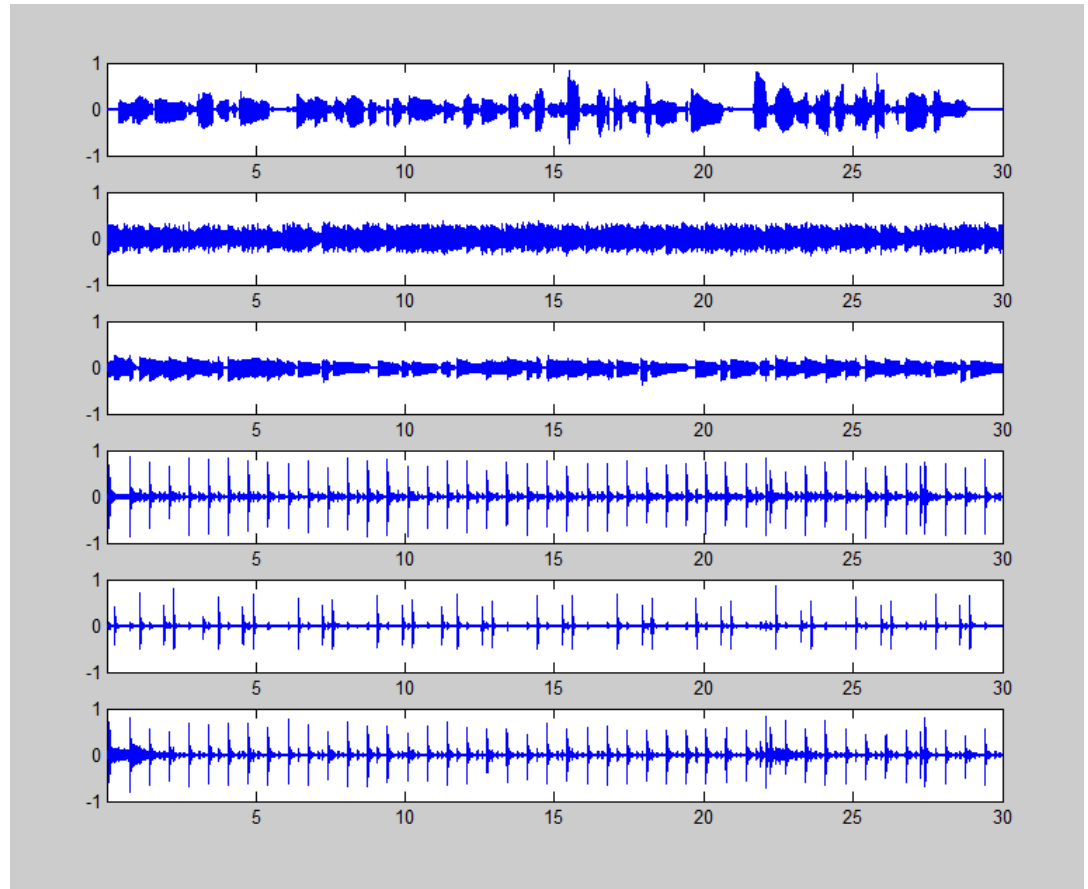
## Loudness normalisation

- Using modified version of BS.1770, to better account for narrowband signals

## For this demo, 6 tracks are used

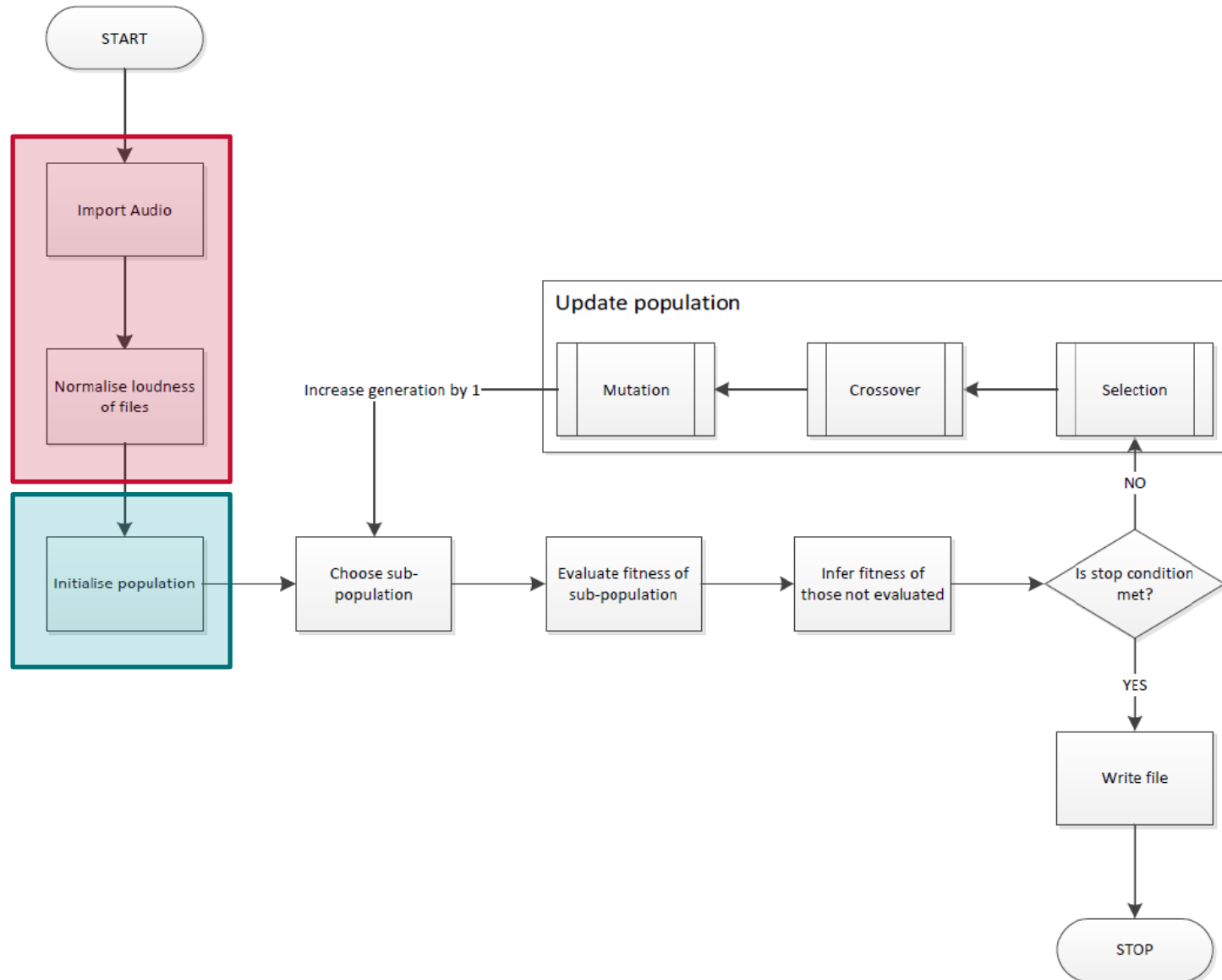
- Vocals
- Guitar
- Bass
- Snare drum
- Kick drum
- Drum overhead

No EQ, Panning, Compression, Reverb, ..., etc.

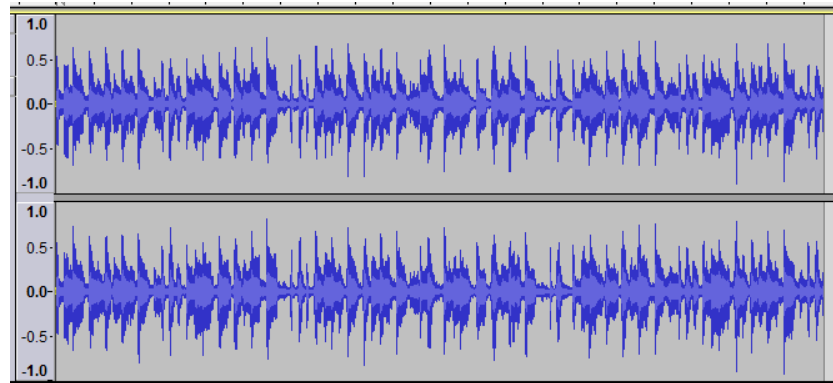
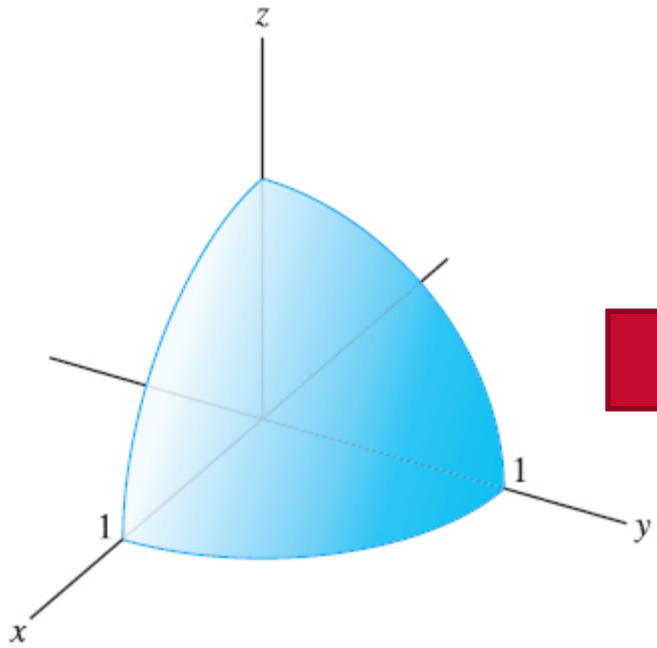


*“Loudness measurement of multitrack audio content using modifications of ITU-R BS. 1770”,  
Pestana et. al., 2013*

# Flowchart





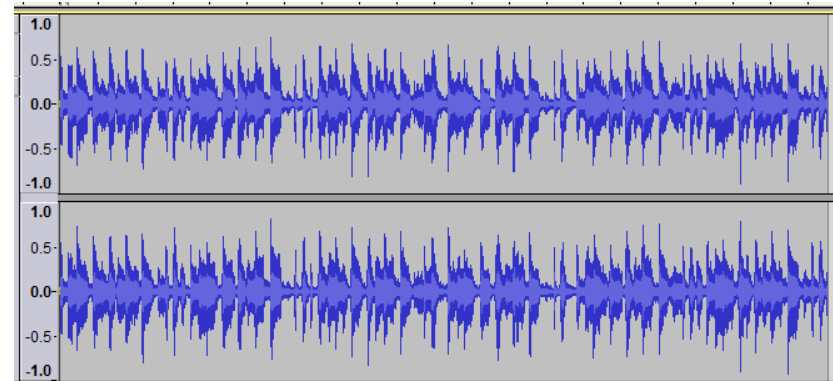
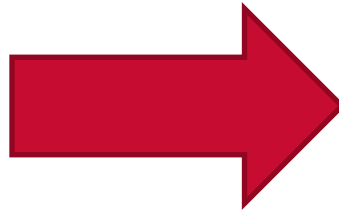


6-tracks, therefore...

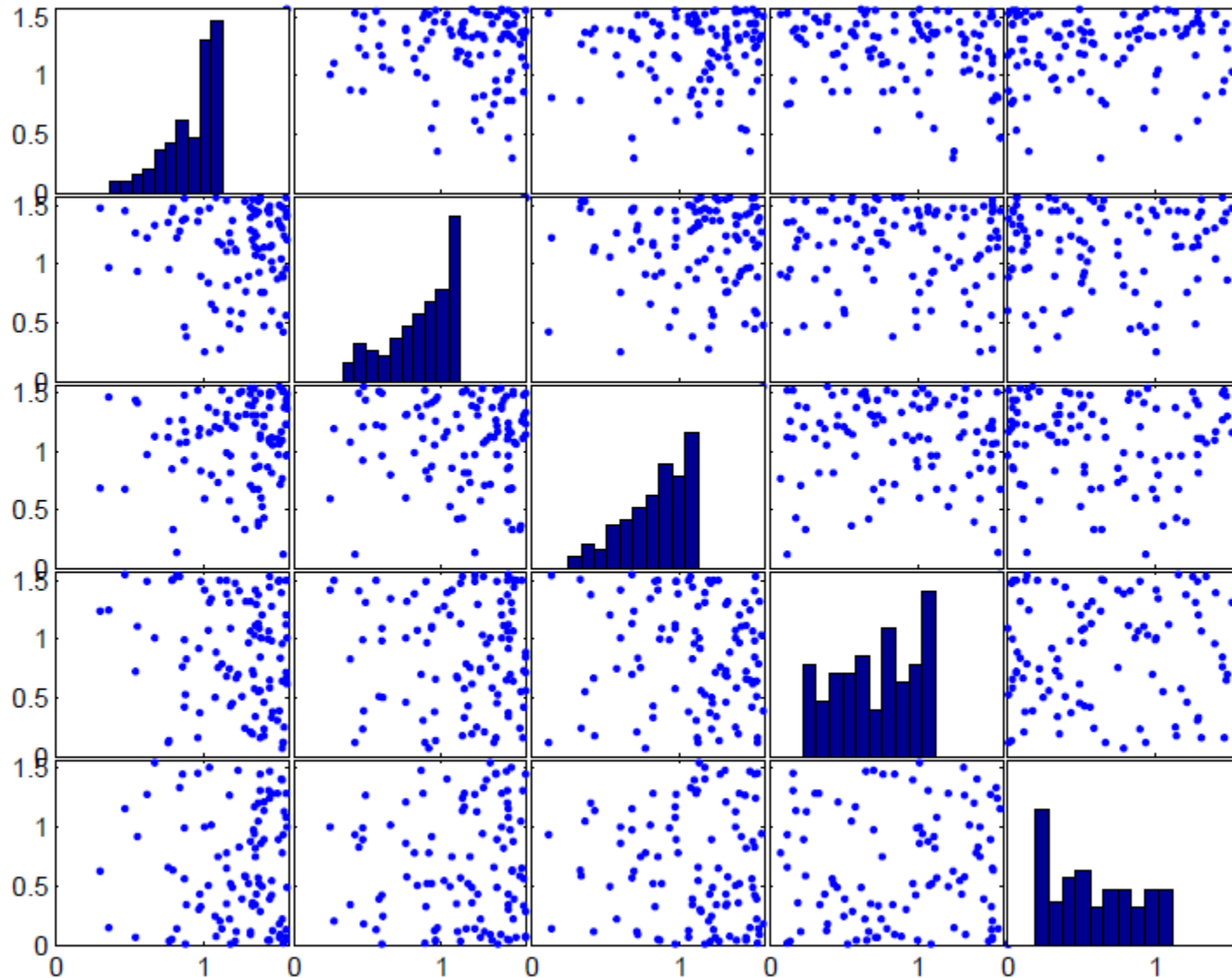
...mixes are on the surface of a sphere in 6-dimensional space...

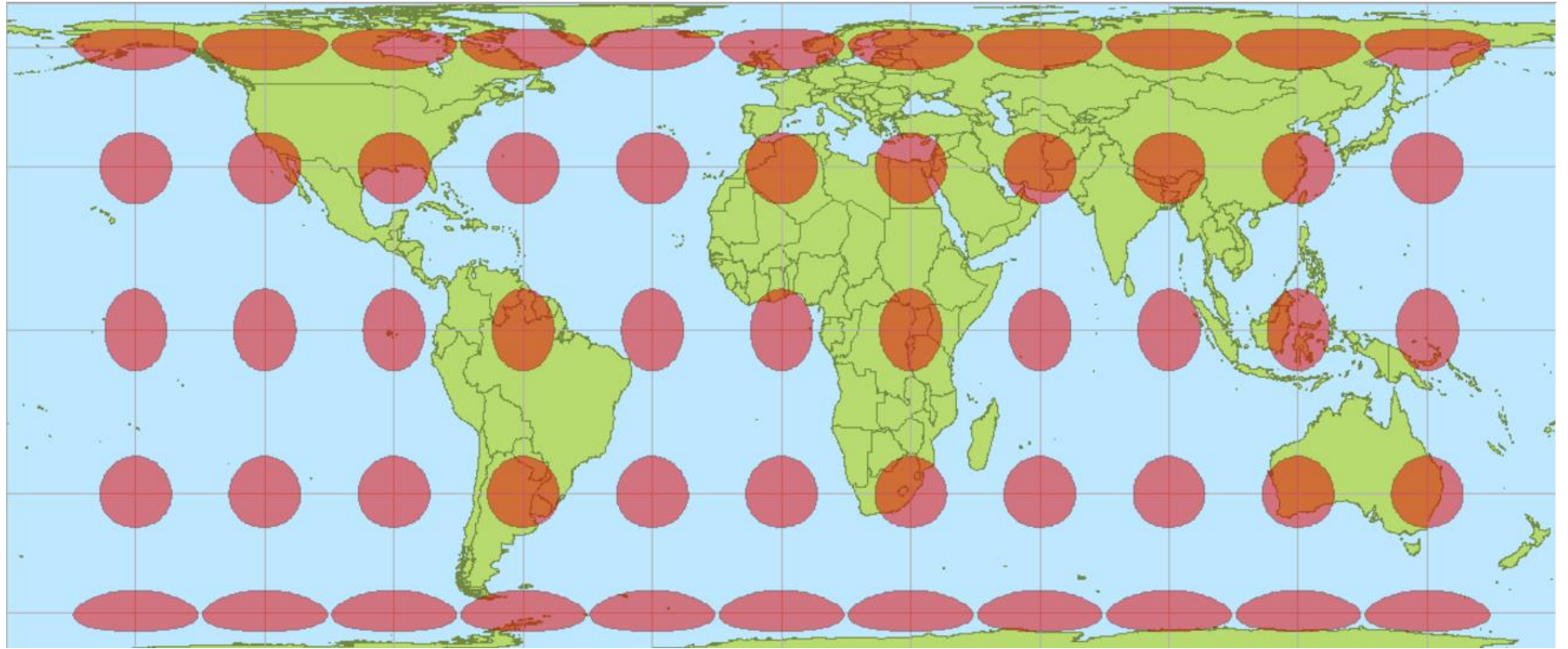
...known as a 5-sphere, as it's surface is 5D.

$\varphi_1$   
 $\varphi_2$   
 $\varphi_3$   
 $\varphi_4$   
 $\varphi_5$

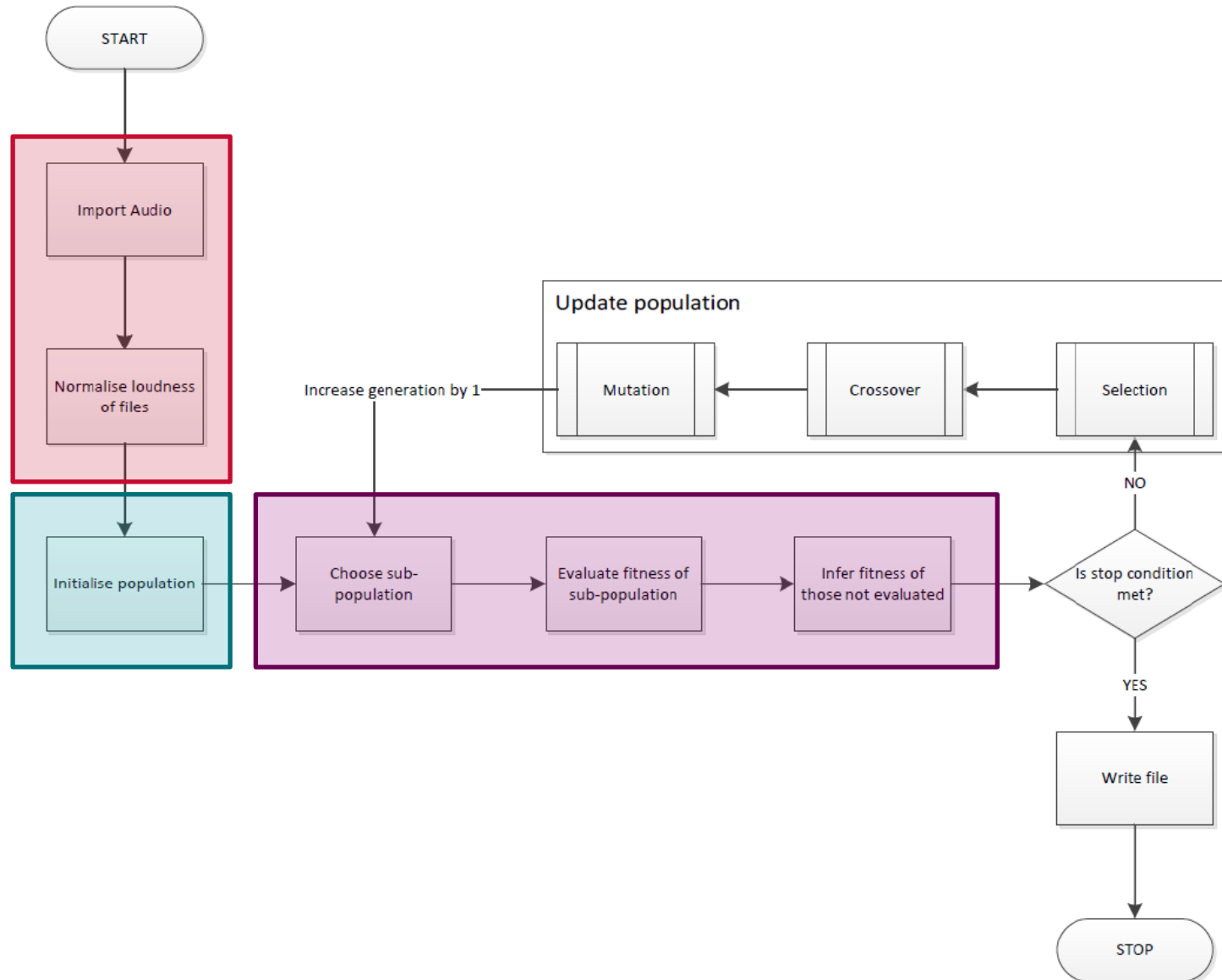


# Initial population





# Flowchart



# Clustering

- Initialise()
- While not end condition do
  - Select()
  - Crossover()
  - Mutate()
  - Do clustering()
  - Pick representatives()
  - Evaluate representatives()
  - Allocate fitness()
- End while

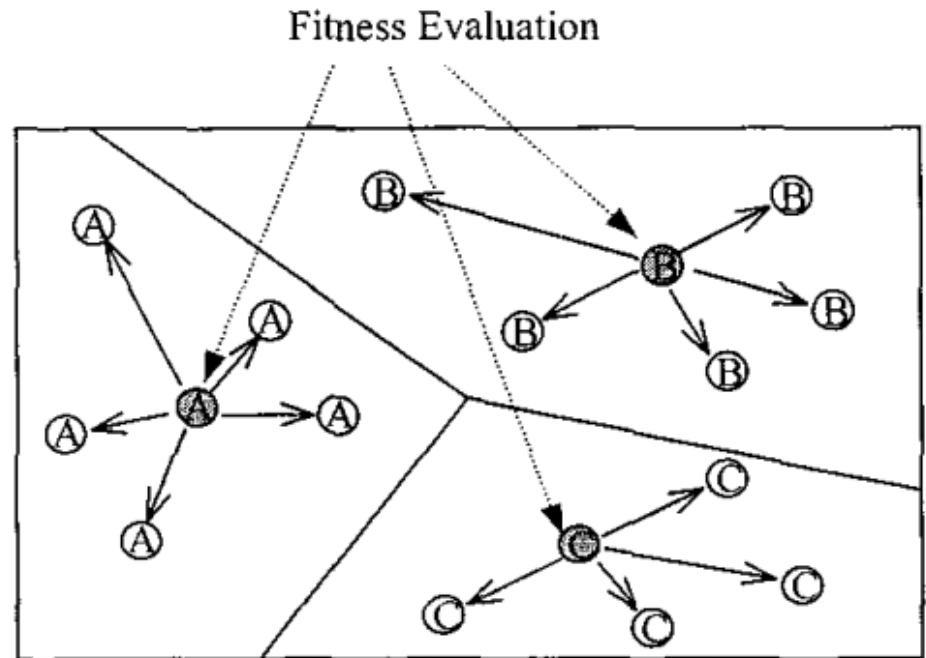
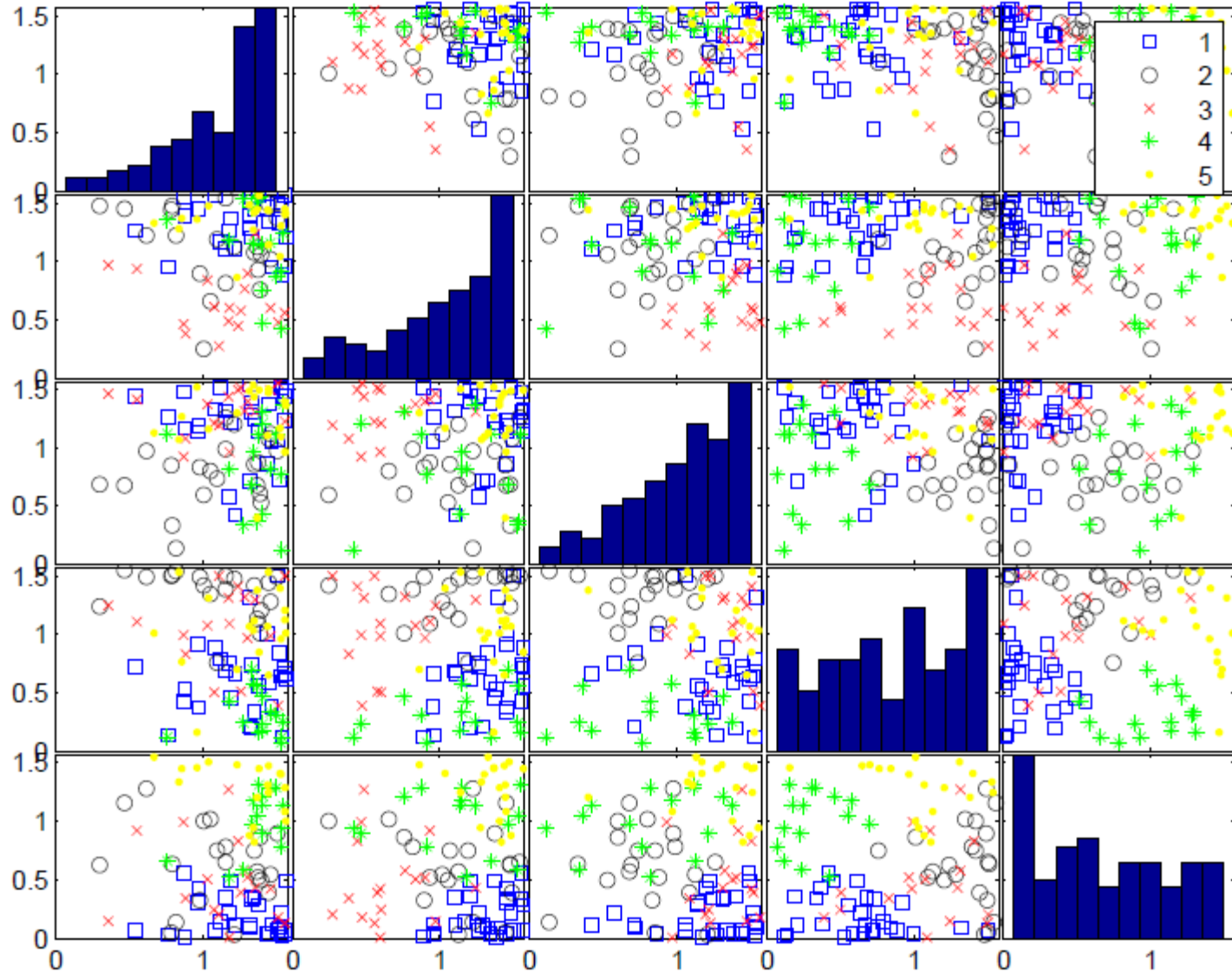


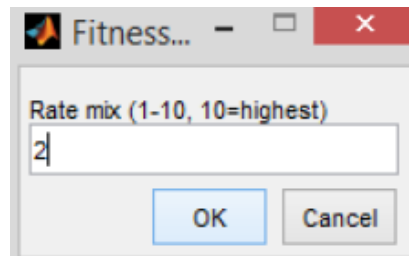
Fig. 6 Sparse fitness evaluation.

# Clustering

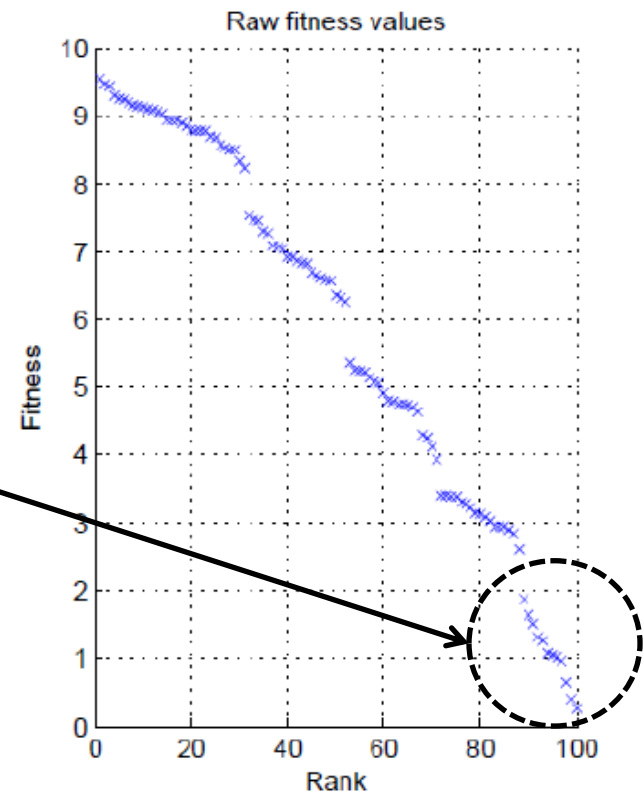


# Evaluate fitness

- Get fitness of representatives
- Assign fitness to individuals not evaluated
- Subjective fitness
  - “How much do you like this mix?”

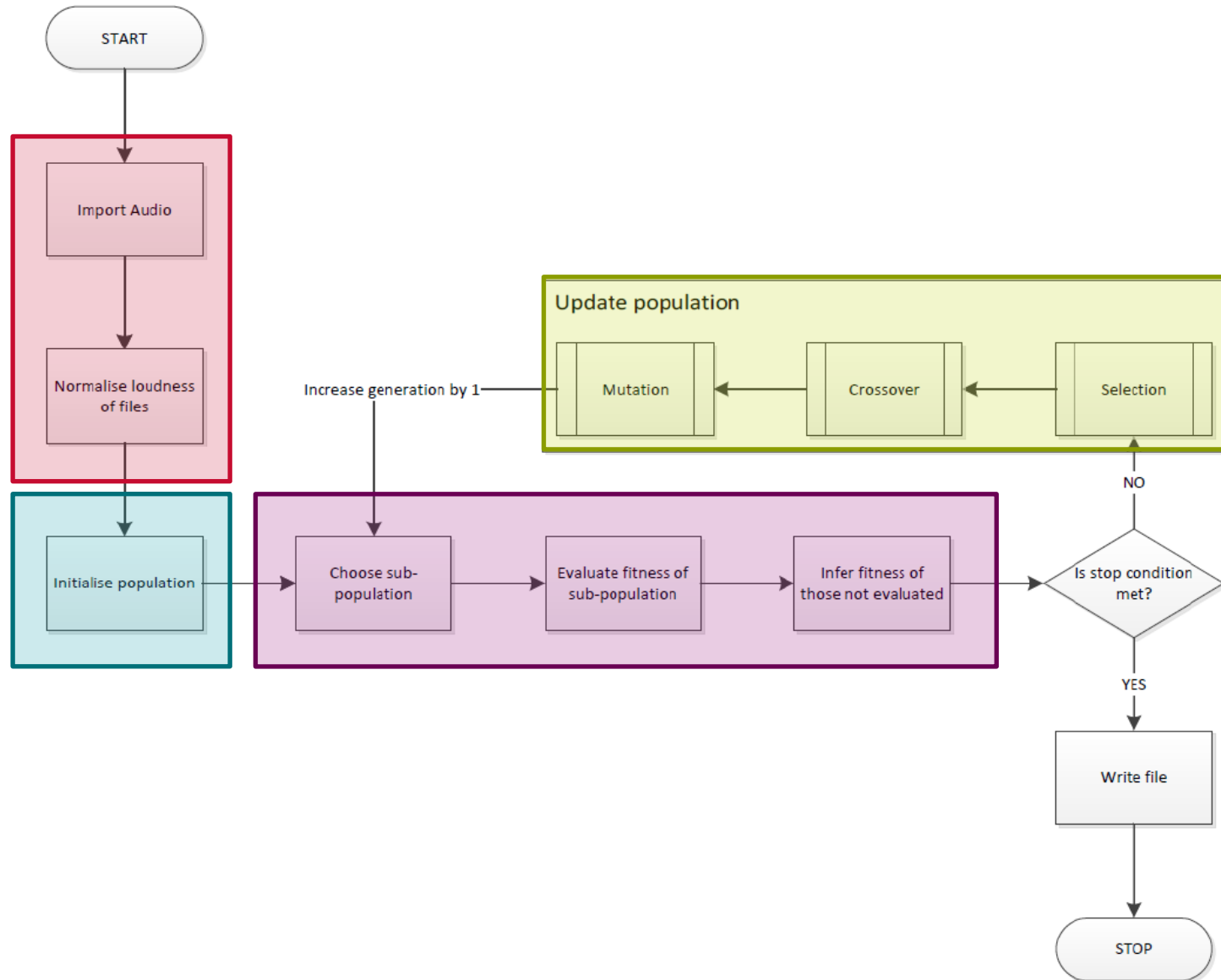


- GA will then maximise this over time



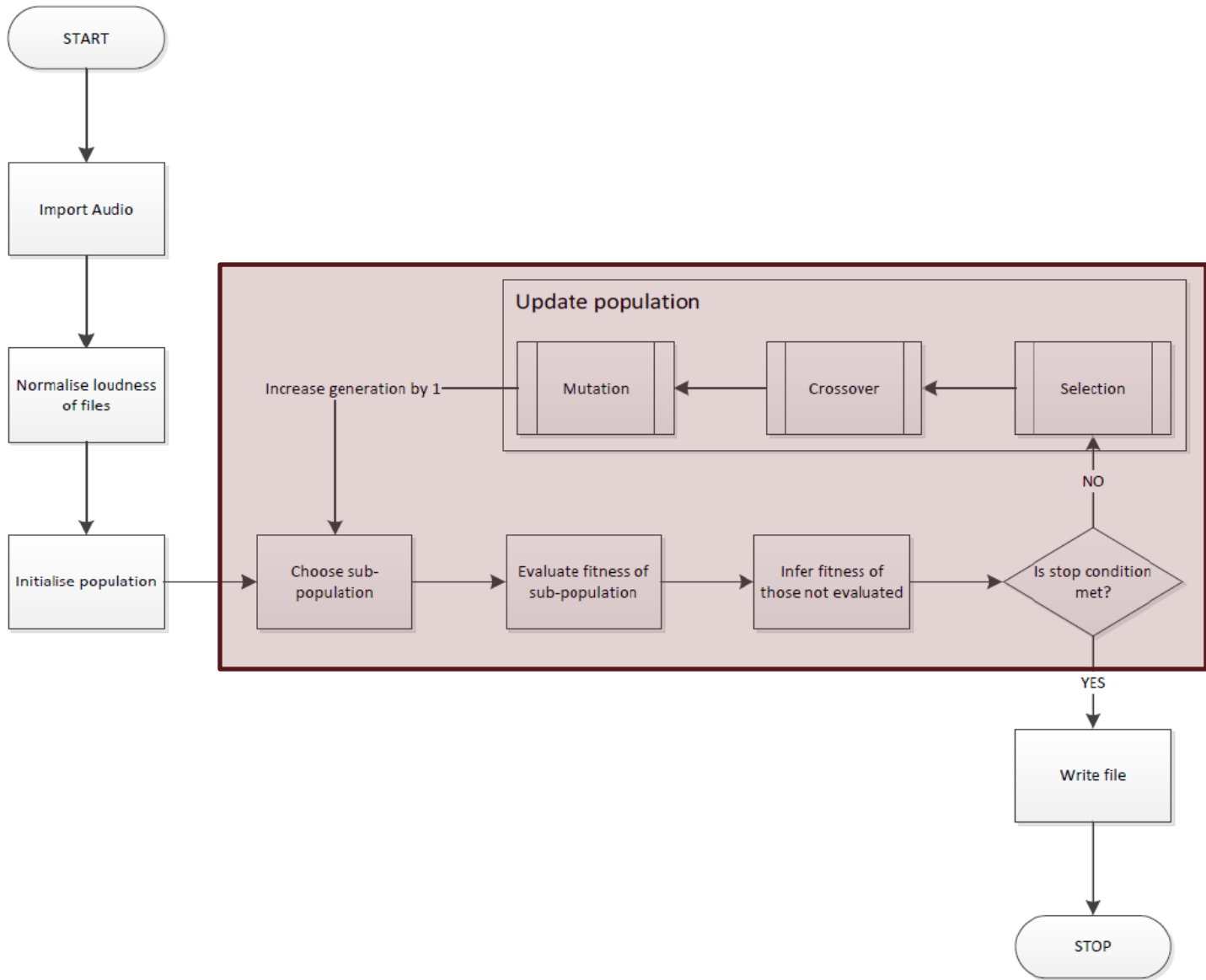


# Flowchart

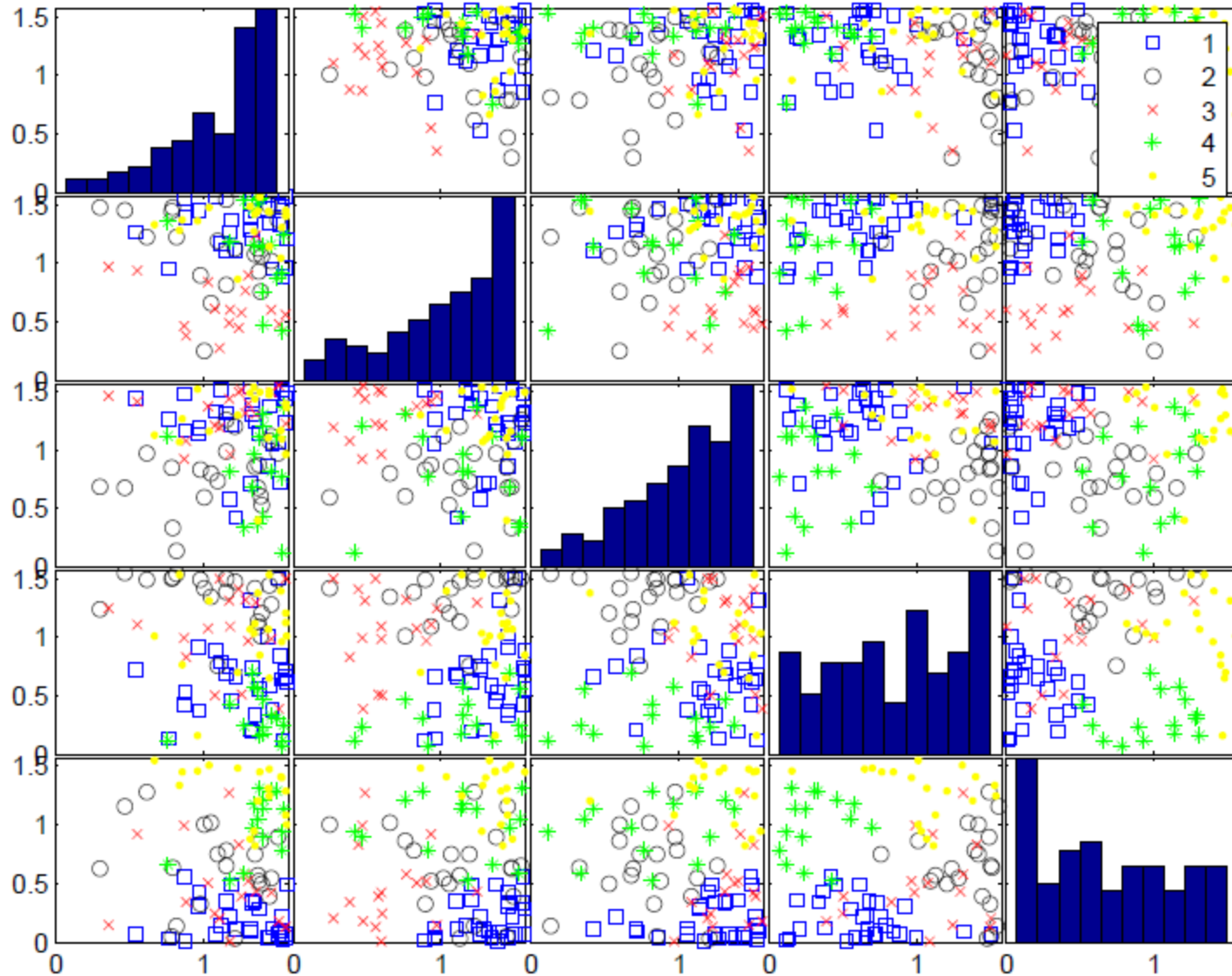


# Update population

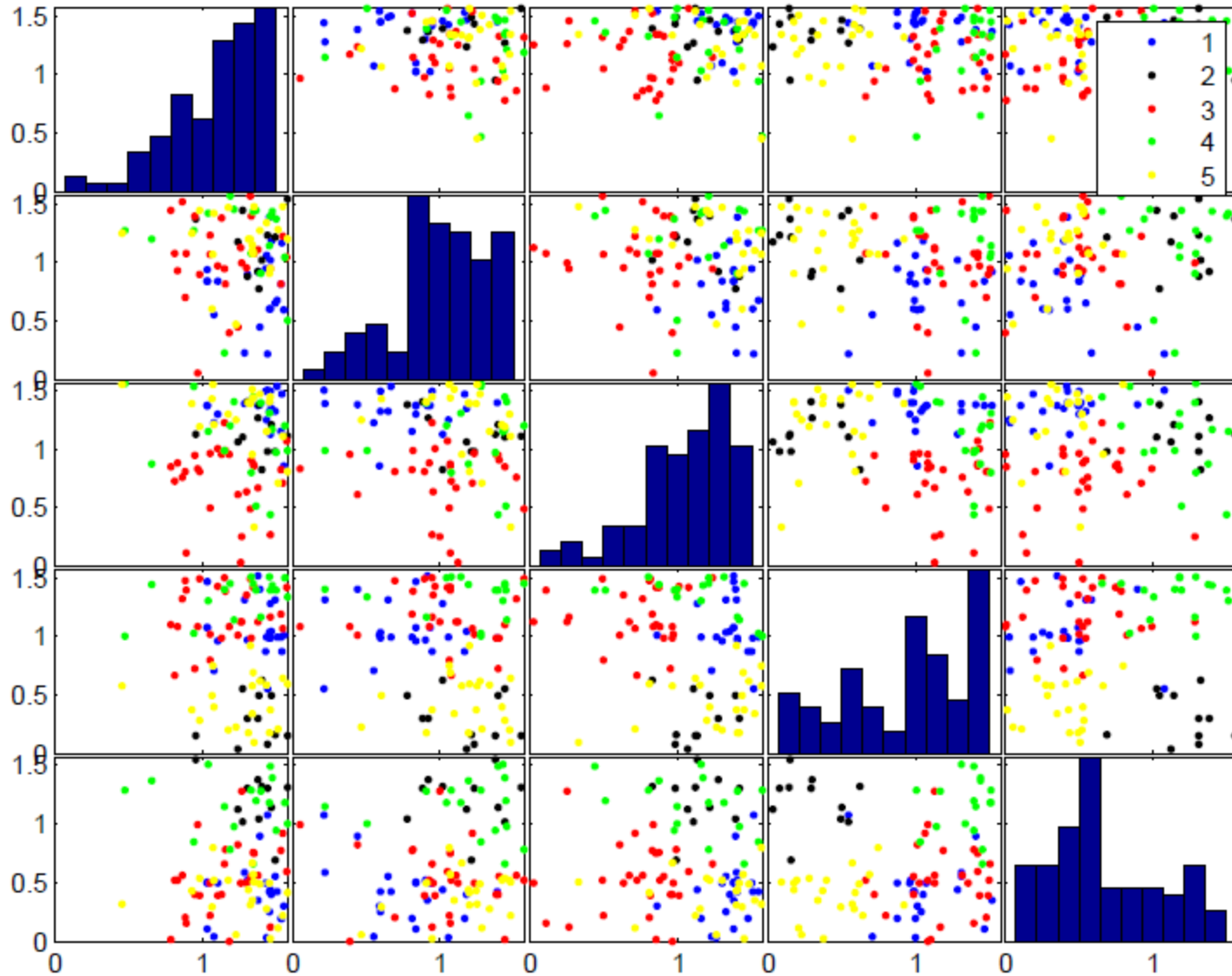
- Fitness Scaling
  - Ranking
- Selection
  - Roulette
- Crossover
  - One-point, **Uniform**, ...
- Mutation
  - Random bit flip



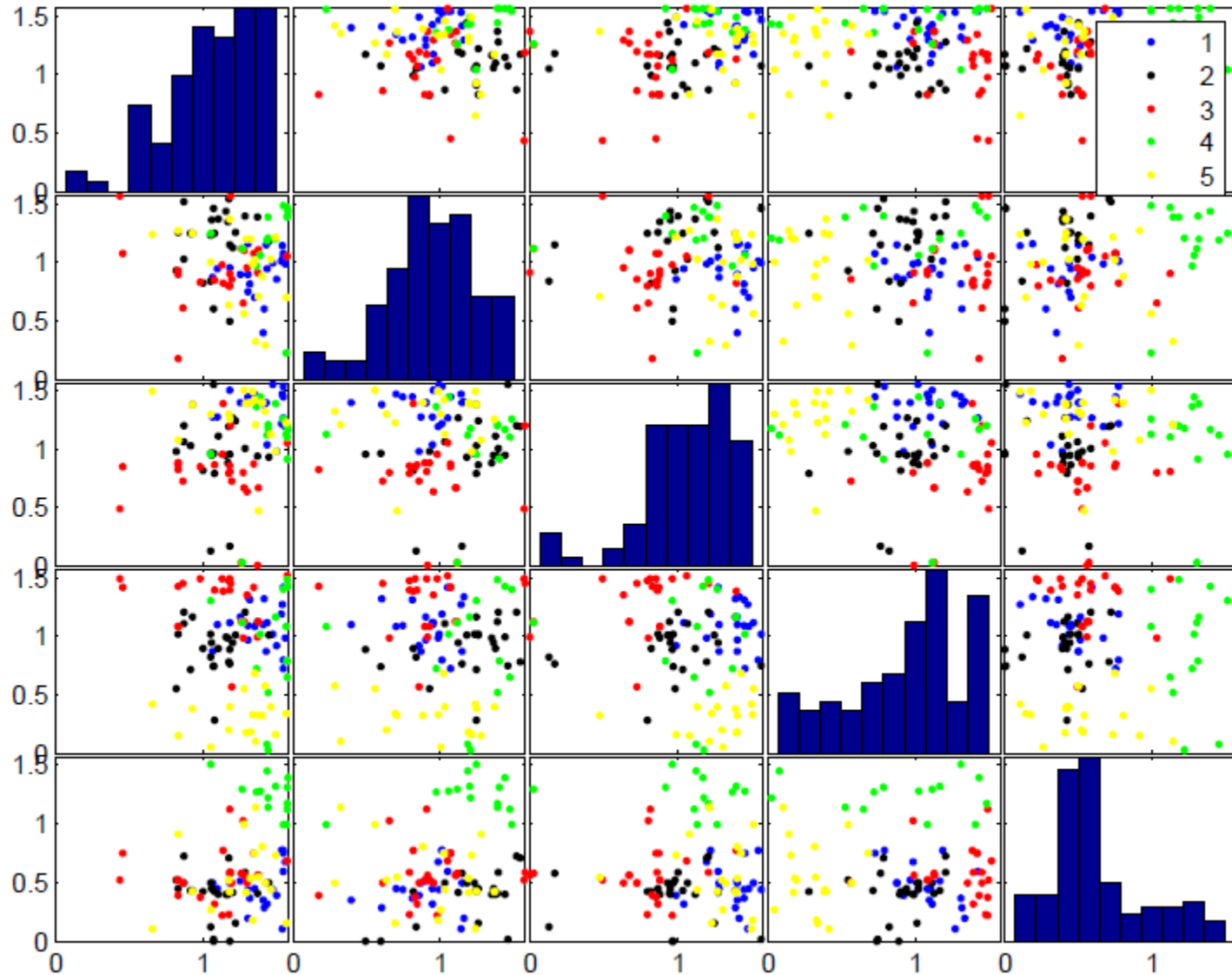
# Initial population



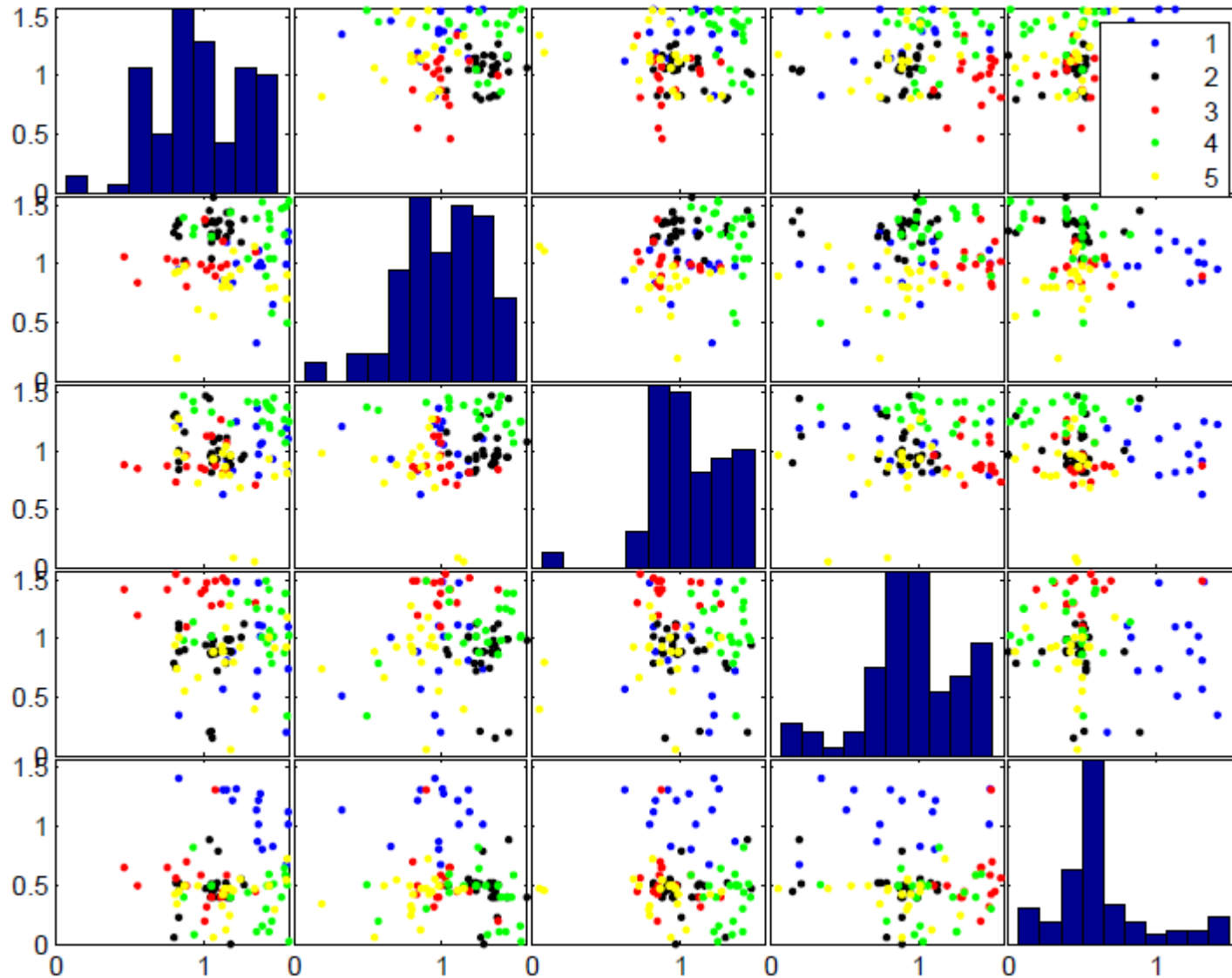
# Gen = 1



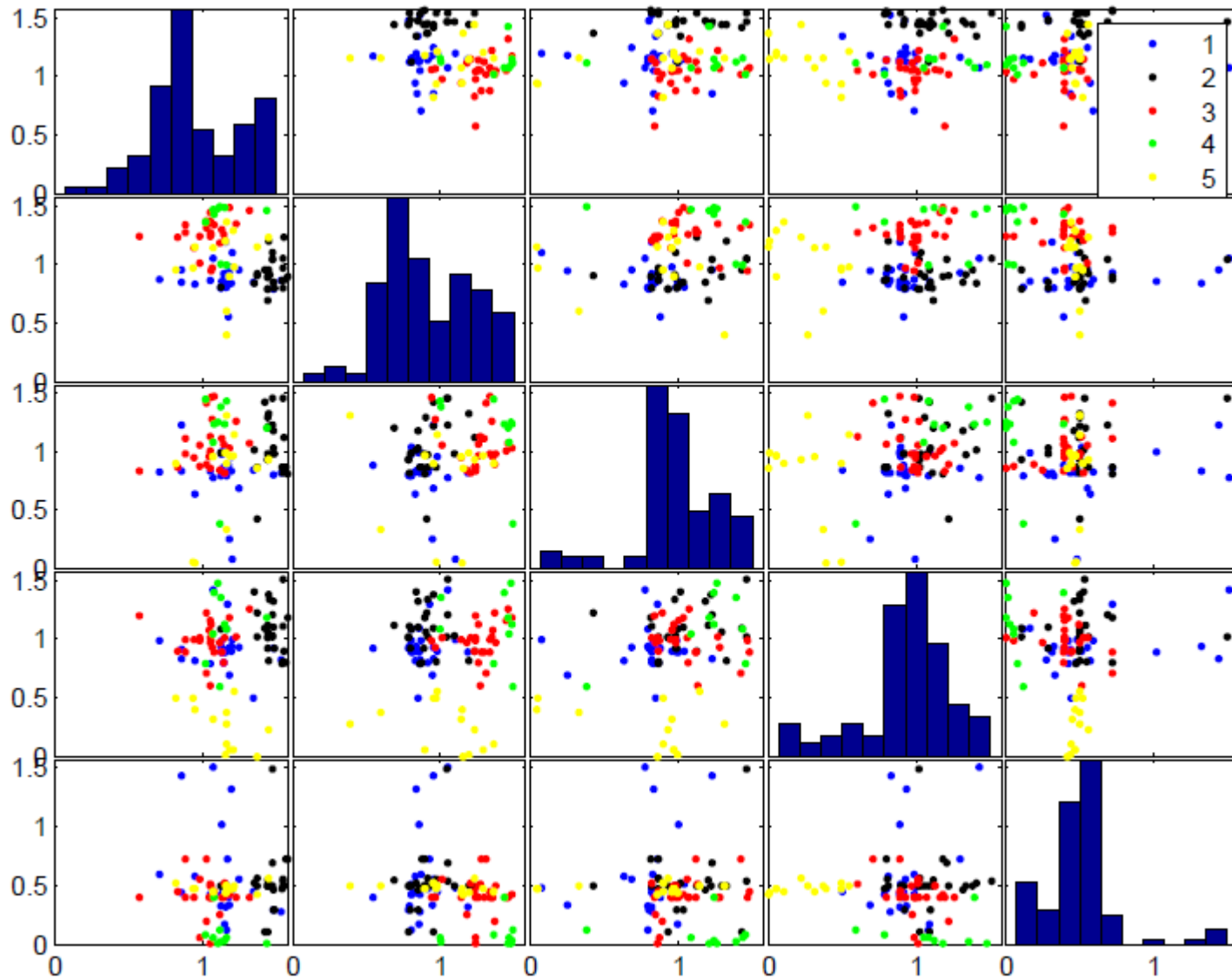
# Gen = 2



# Gen = 3

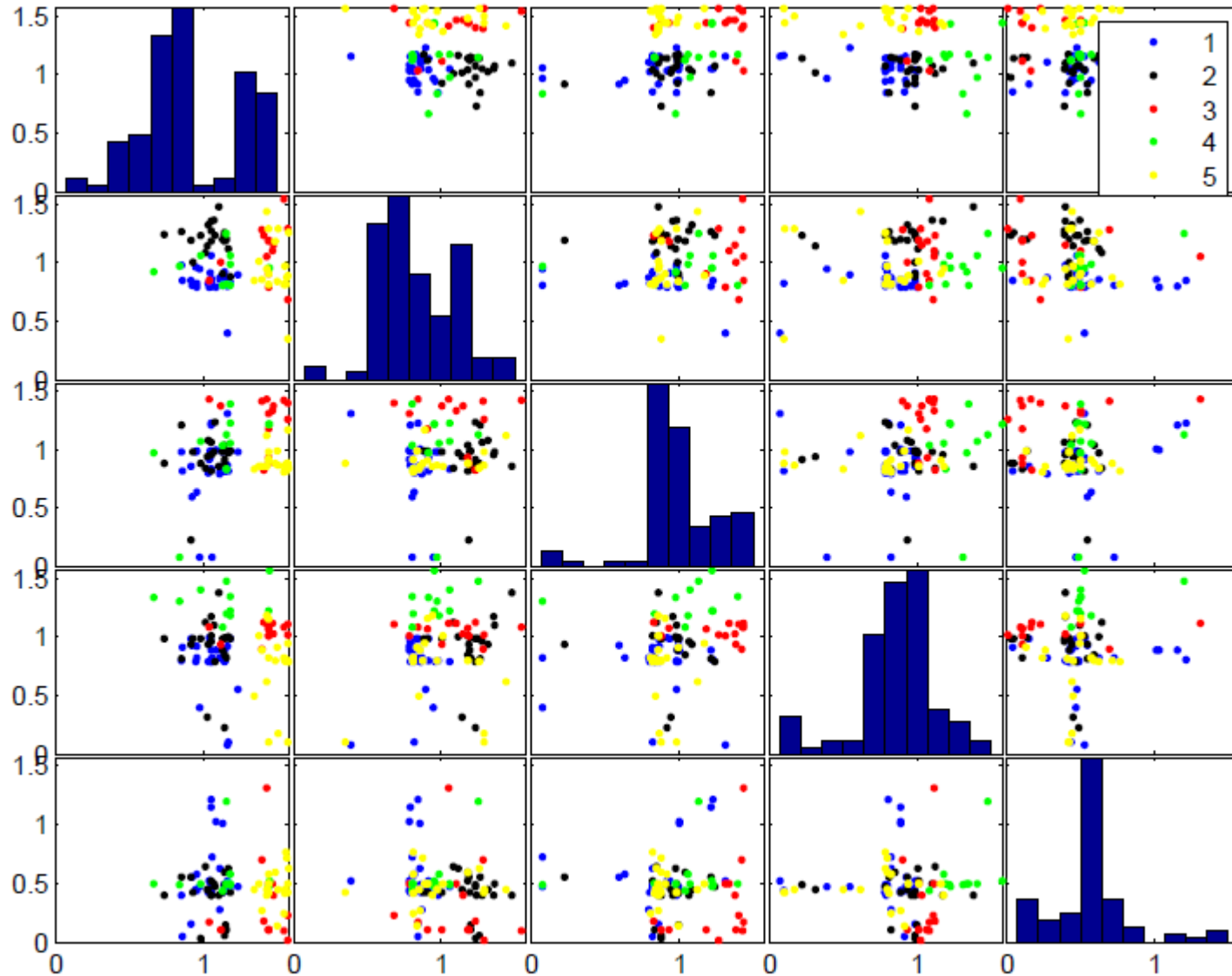


# Gen = 4

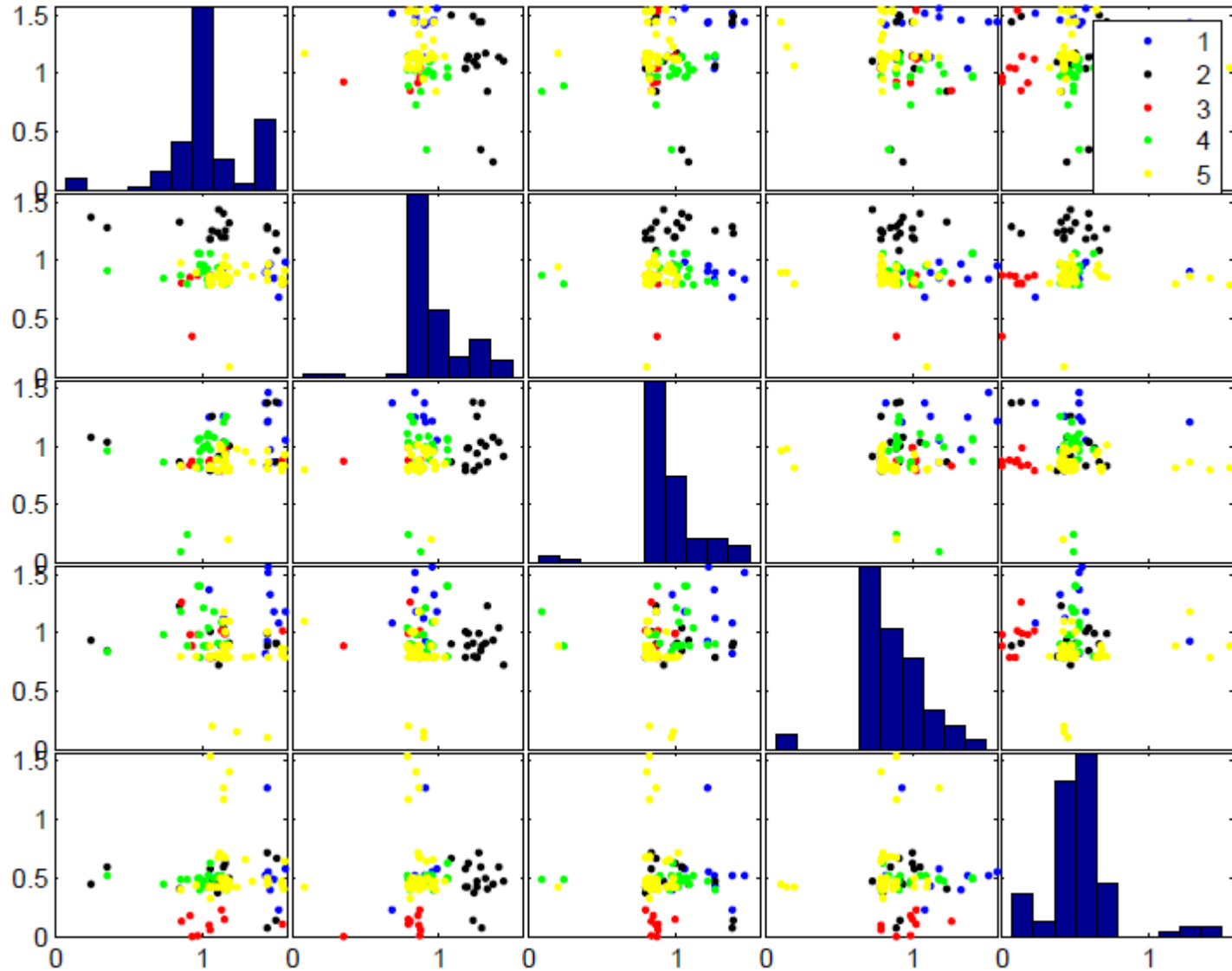




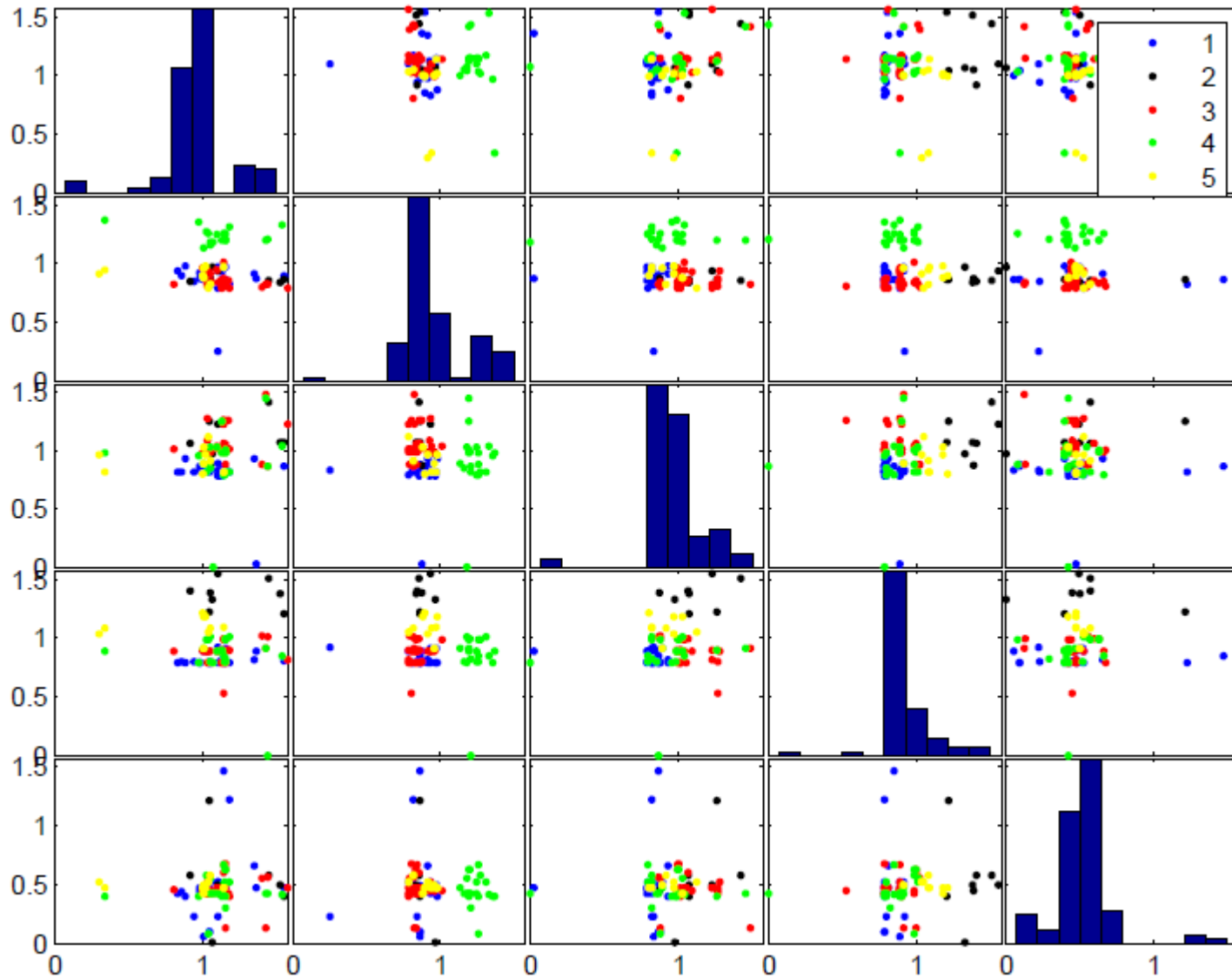
# Gen = 5



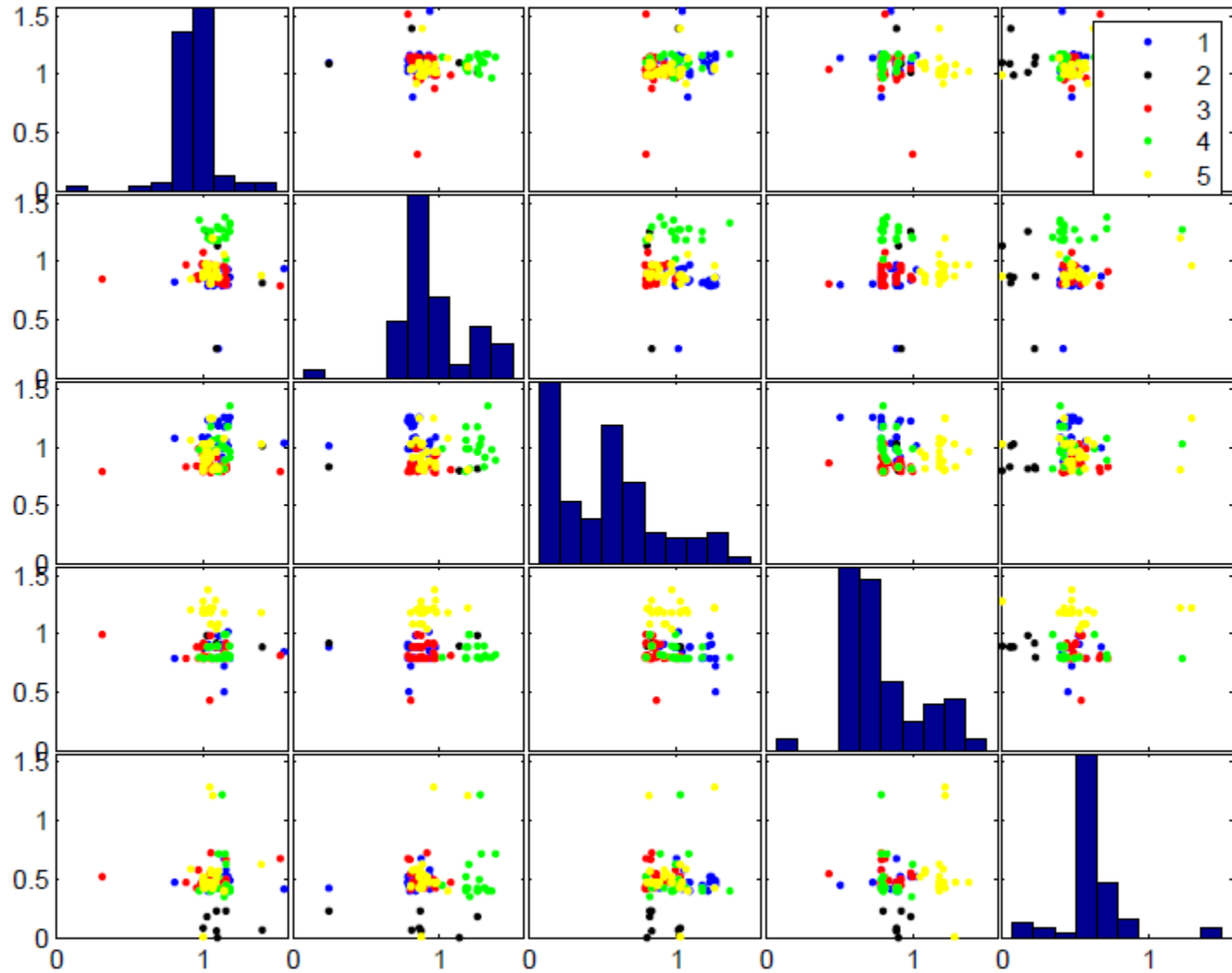
# Gen = 6



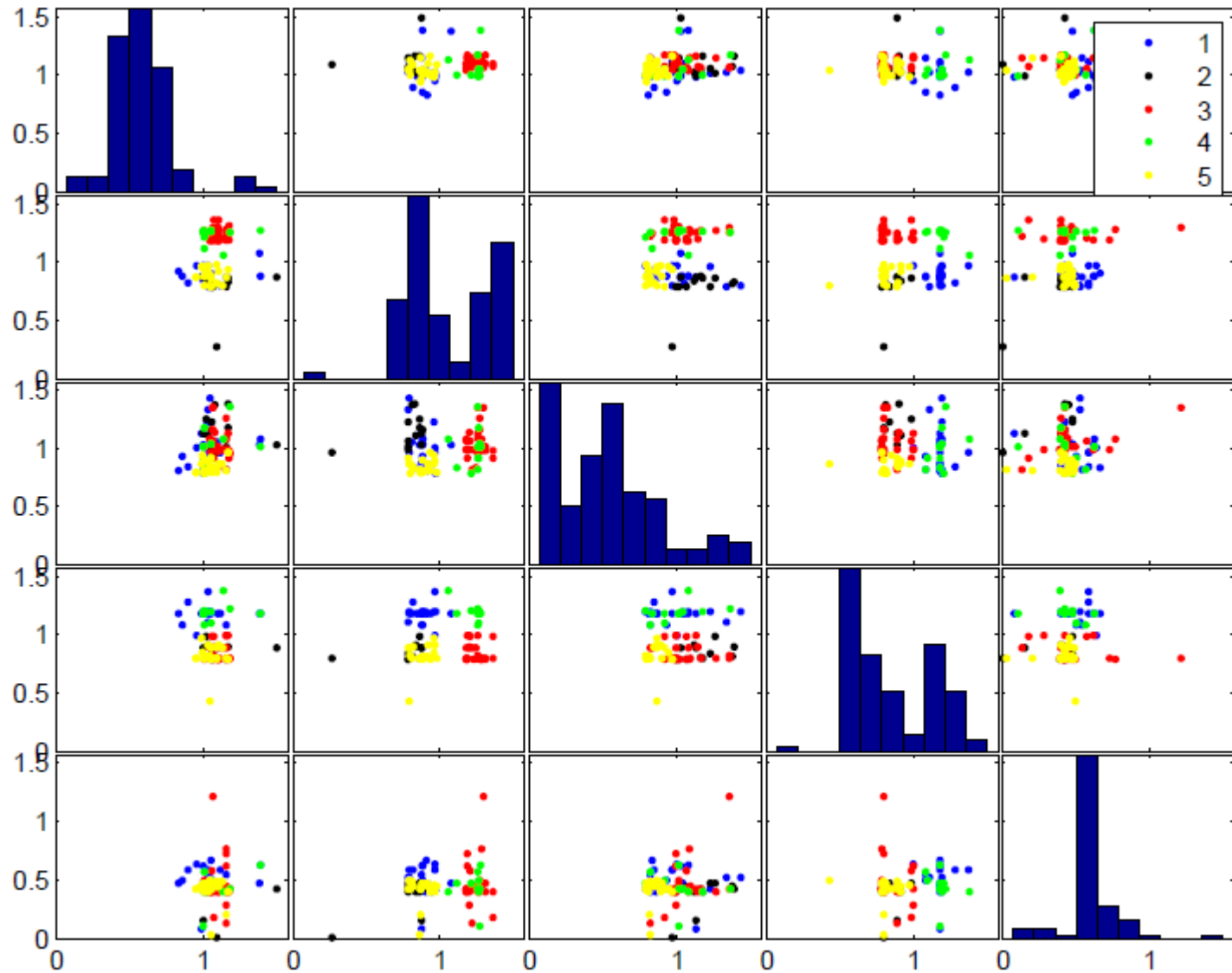
# Gen = 7



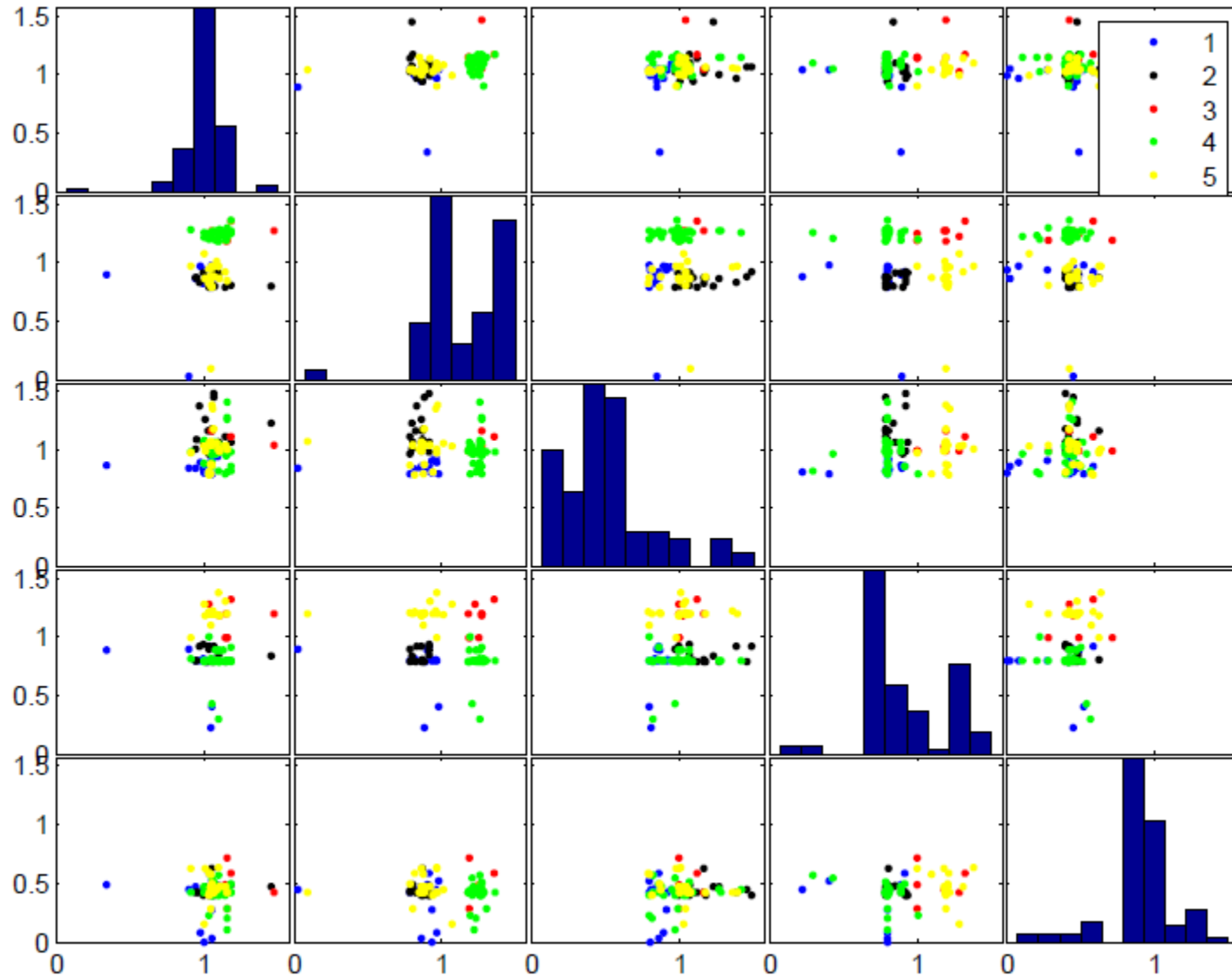
# Gen = 8



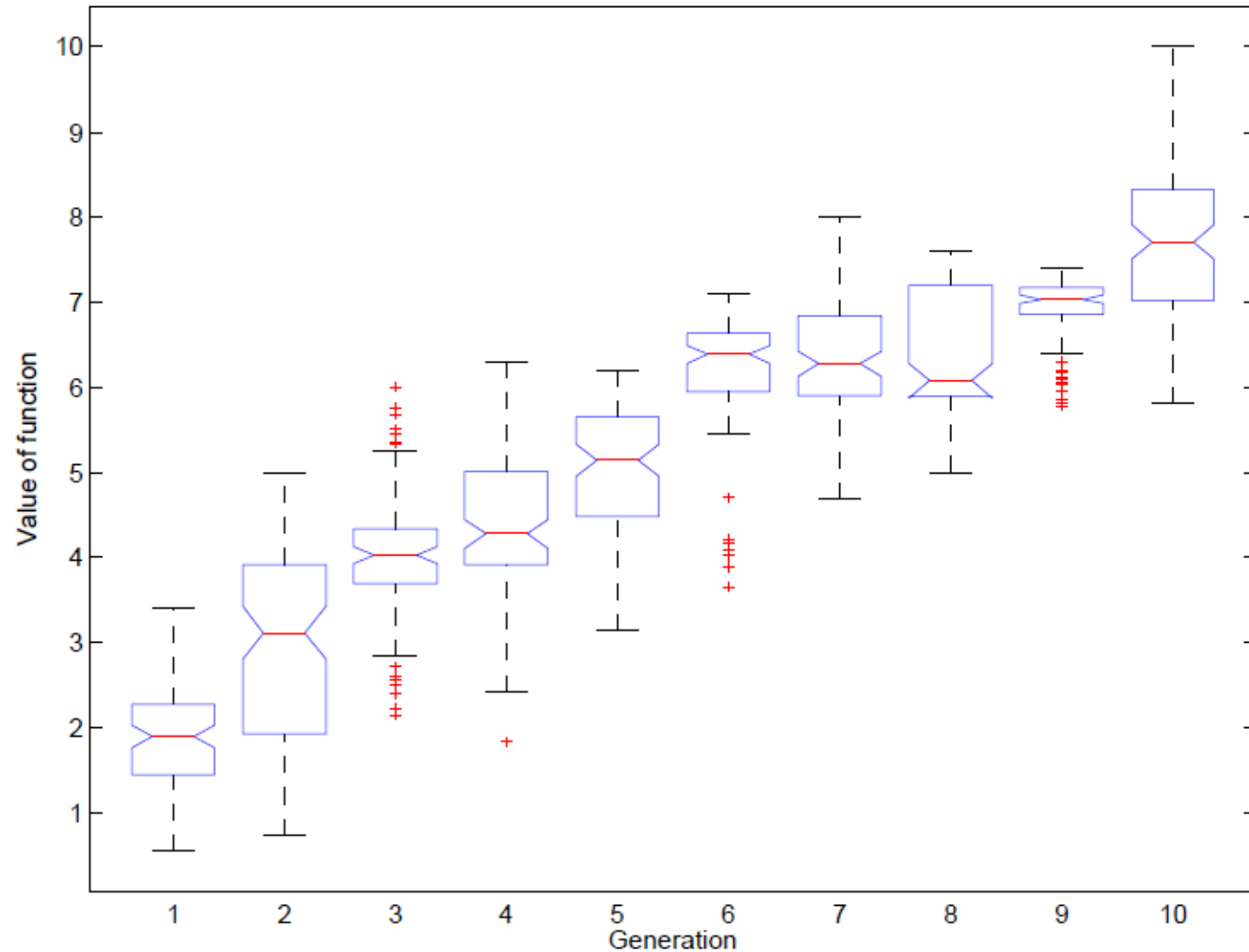
# Gen = 9



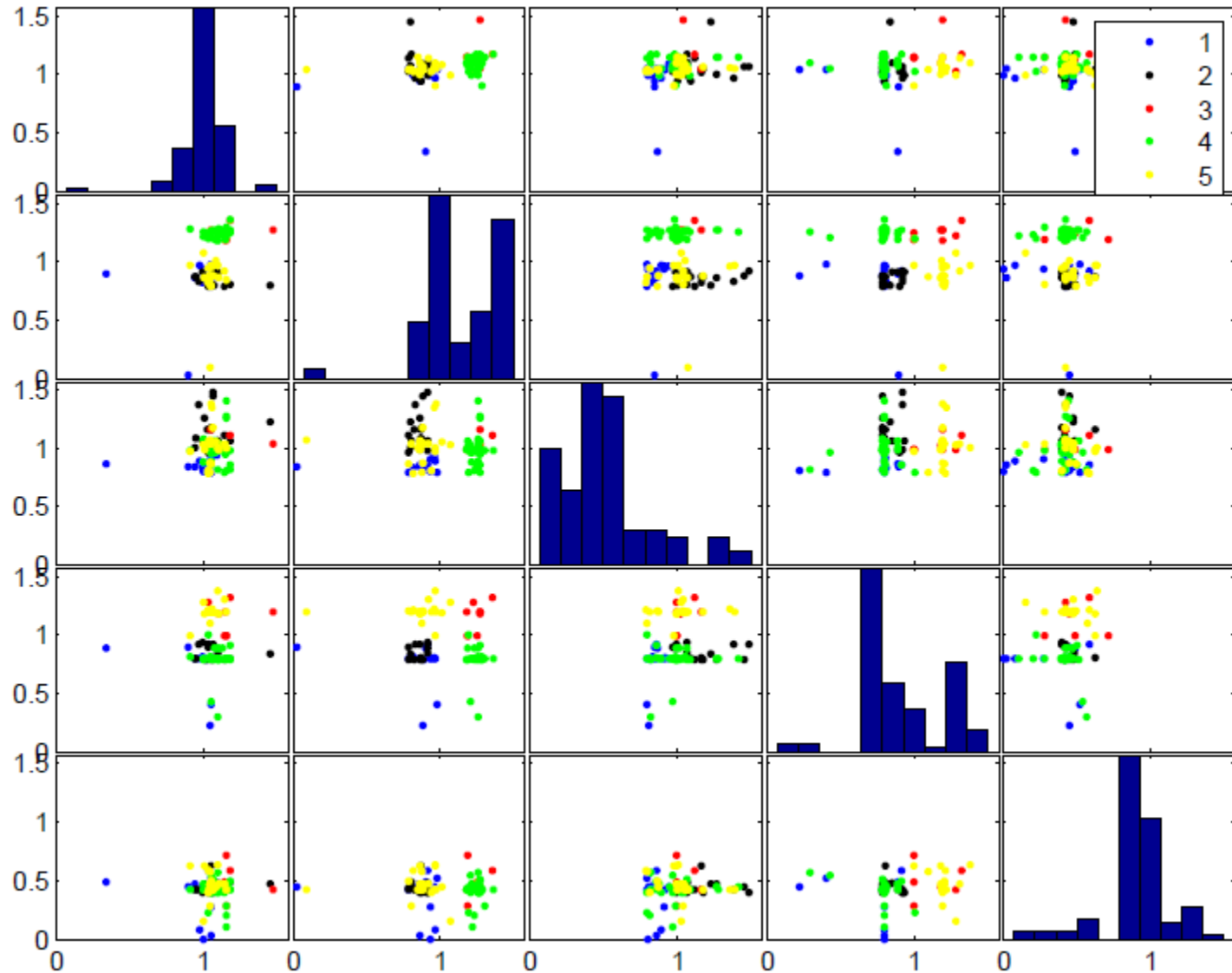
# Gen = 10



# After 10 generations...

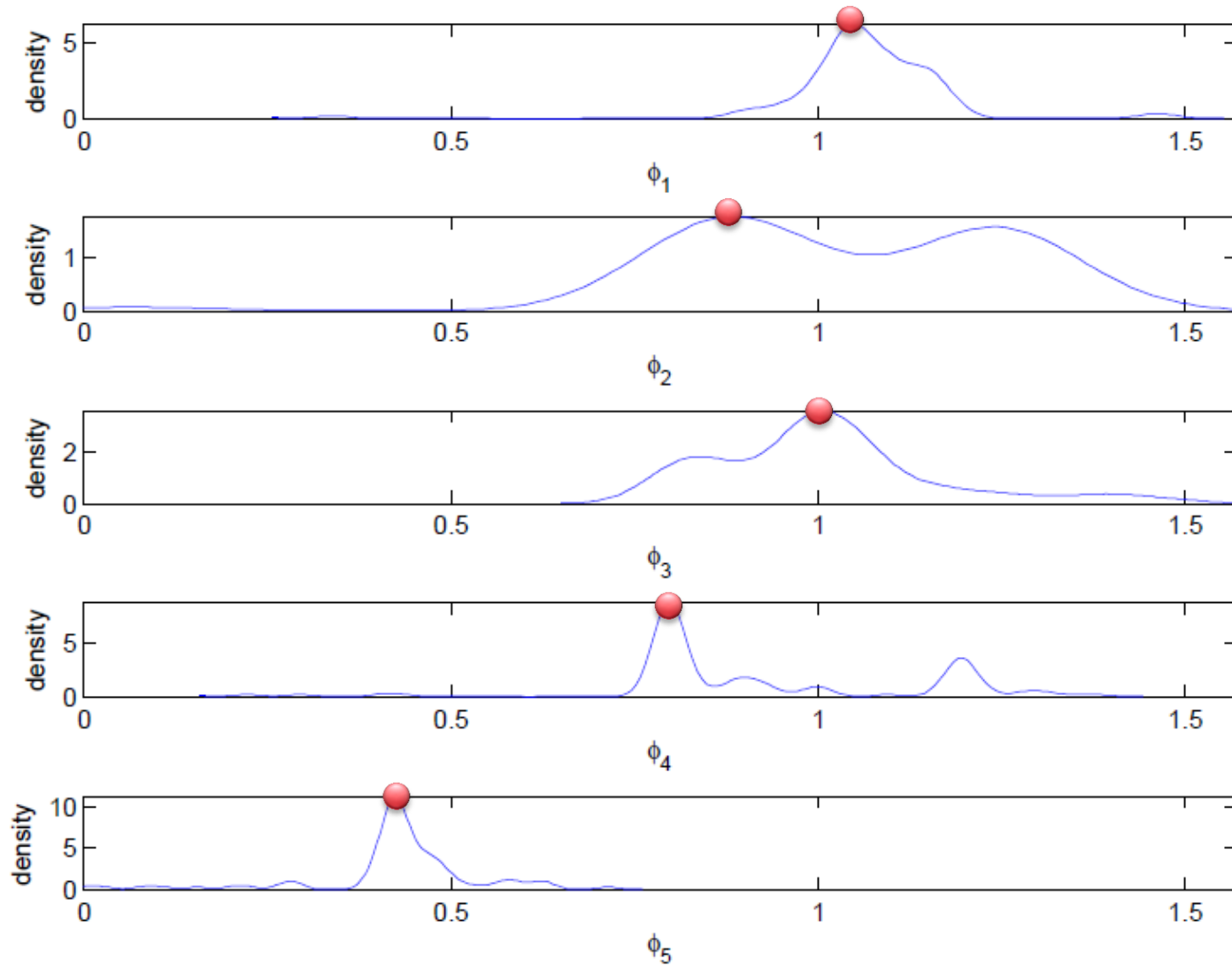


# Gen = 10



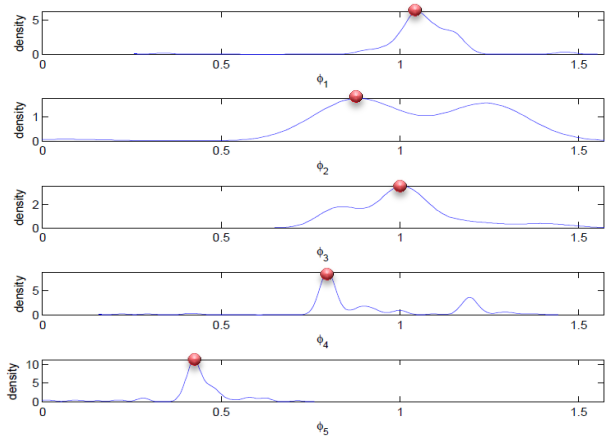


# After 10 generations...



# After 10 generations...

Convert phi terms to gains as follows



$$g_1 = r \cos \phi_1$$

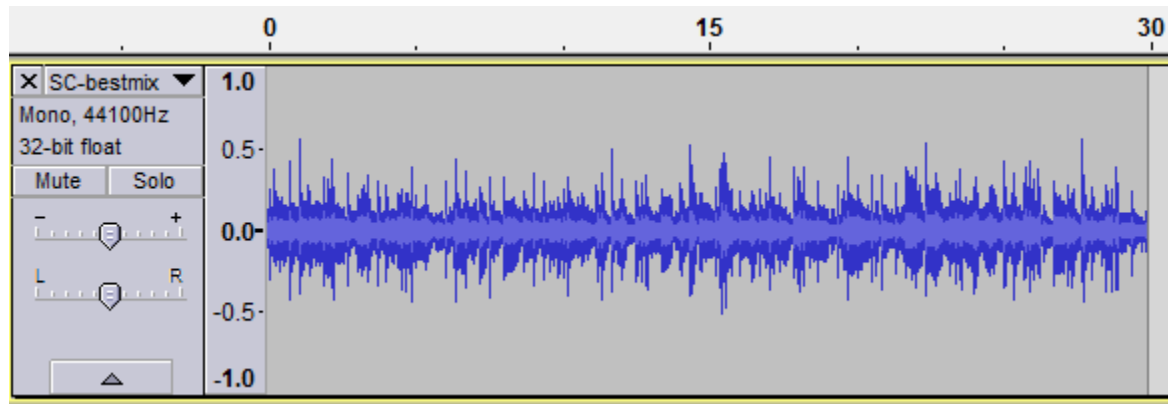
$$g_j = r \cos \phi_j \prod_{k=1}^{j-1} \sin \phi_k$$

$$g_{n-1} = r \sin \theta \prod_{k=1}^{n-2} \sin \phi_k$$

$$g_n = r \cos \theta \prod_{k=1}^{n-2} \sin \phi_k$$

```
mix = audio*gain';
```

```
%generate mix from audio and gains
```

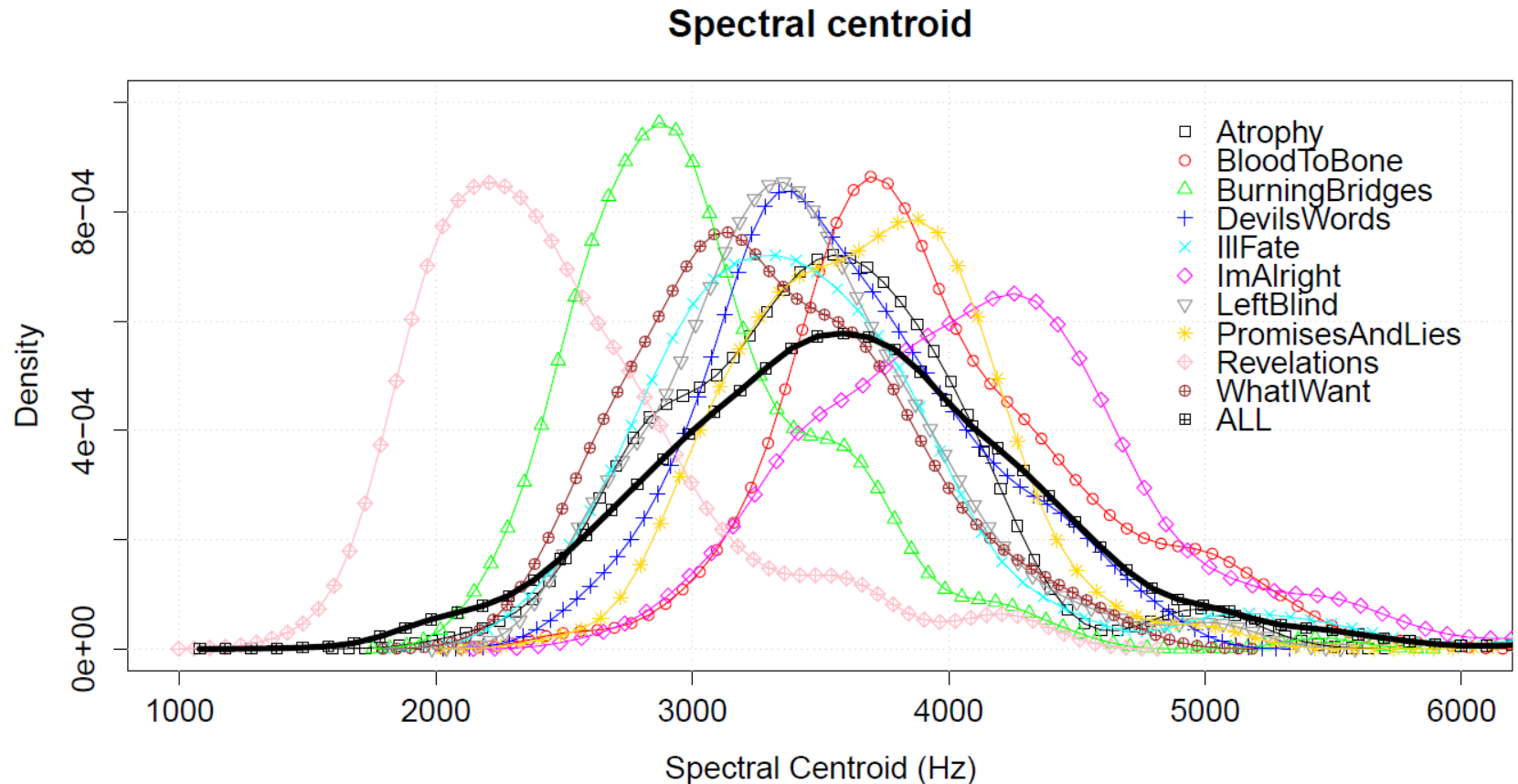


# FURTHER WORK

# Improvements / optimisation

- Optimise
  - Distribution of initial population
  - Number of clusters
  - Selection of representatives from each cluster
- Test the effects of...
  - GA parameters (Population size, Selection, Crossover, Mutation, ..., etc)
- Expand solution space
  - Add equalisation
  - Add stereo panning

# Improvements / optimisation



*“Variation in multitrack mixes: analysis of low-level audio signal features”,  
Wilson/Fazenda, 2016*

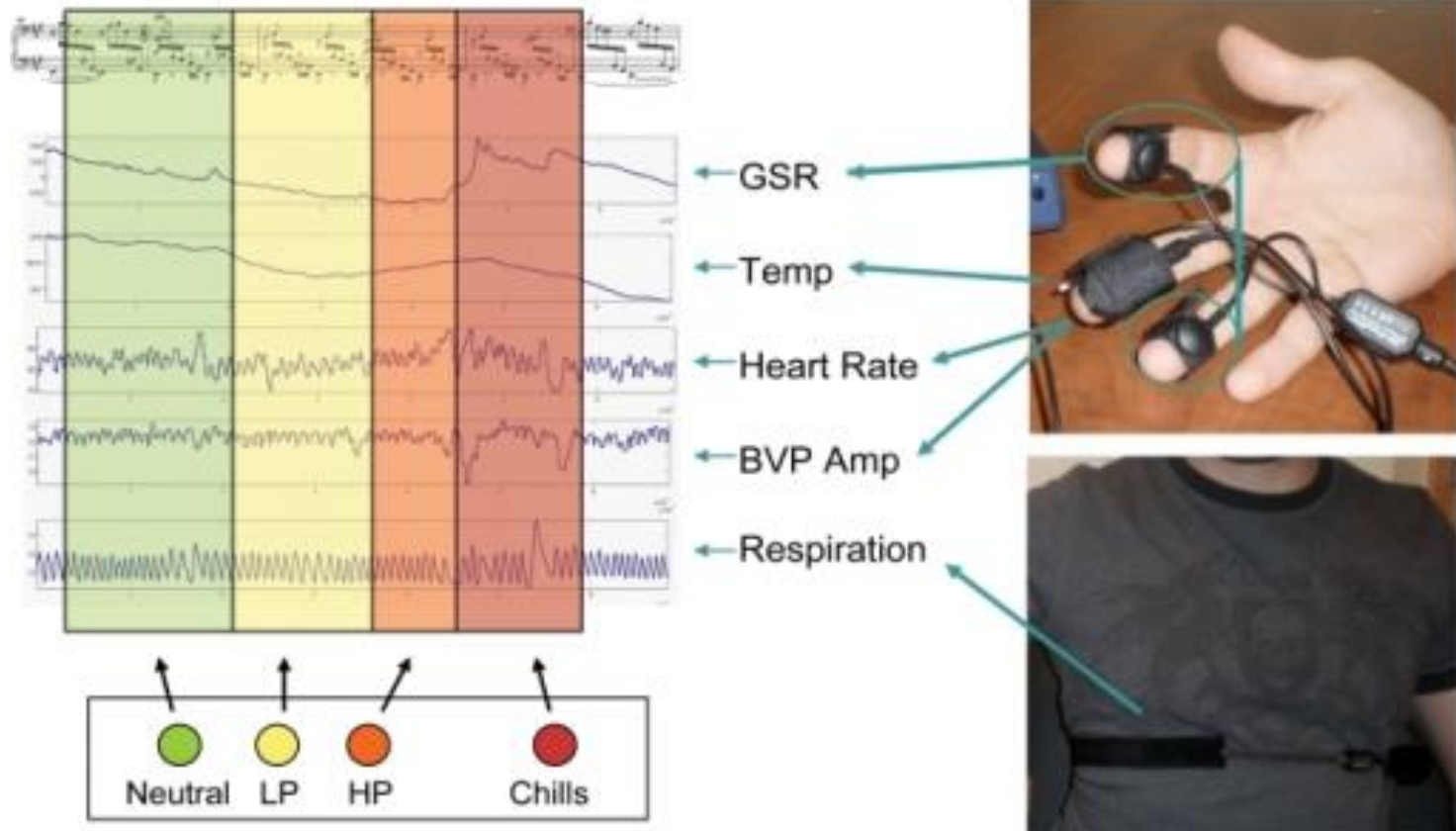
# Fitness function

Subjective (ratings):

- “How much do you like this mix?”
  - Creates mixes you would like
- “How clear is this mix?”
  - Creates mixes that are clear
- “How exciting is this mix?”
  - Creates mixes that are exciting

# Fitness function

Subjective (physiological):



# Conclusions

- Development of system which learns how to present object-based audio according to qualities desired by the user.
- Can be used to learn how certain qualities of audio/music are perceived
- Can be expanded to included further processing



# Thanks for listening

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