

Mosab Mohamed – B20-04 – AI Assignment #2 Report

Task:

- Given a monophonic midi file of a melody. Generate an accompaniment for the melody by using an evolutionary algorithm.
- The accompaniment should be represented by a sequence of chords, each chord should contain three notes.

Solution details:

- We set a fixed length of the chords, so we can have a smaller search space.
 - The fixed length of each chord in the individual is one quarter.
- We have the following representation:
 - Population: list of individuals.
 - Individual: list of chords.
 - Chord: list that consists of 3 notes.
 - Note: an integer number that represents the midi value of a note.
- We want an accompaniment that compliments the original melody. And to achieve that we need to:
 - Be similar to the original melody in terms of flow.
 - Not be in the same octave as the original melody to avoid dissonance.
 - Have the chords follow the original melody's key.
 - Avoid having the same chord played consecutively for too long.
- To ensure progress in each generation:
 - We keep the best 50% of the previous generation.
 - We produce the other 50% of the population from the best 50% of the previous generation.
- To ensure variation in each generation:
 - We have a relatively high mutation probability which is 5% for each chord in the individual.
 - We mutate by replacing a chord in the individual with an entirely new random chord.

Algorithm Description:

- The program takes an input midi file (*.mid) that consists of melody and applies genetic programming(evolutionary algorithm) to generate an accompaniment consisting of several three note chords(triads) that would sound pleasing while played with the input melody.

Dependencies

- Python 3.8
- Python library [mido](#)
 - Used to parse the input midi file to readable values of notes and times.
 - Used to write the output file using the tracks of the input and the generated accompaniment.
- Python library [music21](#)
 - Used to detect the key of the input melody to help us in the fitness function.
- Python library [numpy](#) and [random](#)
 - Used to help us execute various random operations.

Genetic algorithm components:

- **Representation:**
 - The **population** is represented as a list of possible **individuals**(accompaniments).
 - Each **individual** is a list of N **chords**, the individual represents a possible solution to the problem.
 - Each **chord** is a list of 3 **notes**, the chord represents a chord from the following set of chords:
 - Major triad, Minor triad.
 - First inverted major triad, First inverted minor triad.
 - Second inverted major triad, Second inverted minor triad.
 - Diminished chord.
 - Suspended second chord.
 - Suspended fourth chord.
 - Rest.
 - Each **note** is an integer number that represents the midi value of a specific note.
 - Each **chord** object is played for one quarter duration which equals 384 in the mido library time units.

- **Fitness function:**

- The fitness value of an individual depends on three aspects:
 - The similarity of each chord in the individual to the average note played in the same quarter of the original melody in a lower octave.
 - While for empty quarters we calculate the similarity to the average note played in the next quarter that does play some notes.
 - To avoid dissonance we change the target average we want to reach to the same melody but shifted 2 octaves below the original melody.
 - For example if we have an average note played in a quarter of the original melody that is equal to C5.

We make the targeted average be C3.

- The existence of each chord of the individual in the possible chords we can compose out of the original melody key.
 - Based on the tables presented in the assignment, and that a valid chord of some key is three notes that start from a root note in the table and between a pair of consecutive notes we skip a note.
e.g : I-iii-V in major, i-III-v in minor.

Minor Keys	i	ii°	III	iv	v	VI	VII	Major Keys	I	ii	iii	IV	V	vi	vii°
Cm	Cm	D°	E♭	Fm	Gm	A♭	B♭	C	C	Dm	Em	F	G	Am	B°
C#m	C#m	D#°	E	F#m	G#m	A	B	C#	C#	D#m	E#m	F#	G#	A#m	B#°
Dm	Dm	E°	F	Gm	Am	B♭	C	D	D	Em	F#m	G	A	Bm	C#°
D#m	D#m	E#°	F#	G#m	A#m	B	C#	D#	D#	Em	F#m	G	A	Bm	C#°
E♭m	E♭m	F°	G♭	A♭m	B♭m	C♭	D♭	E♭	E♭	Fm	Gm	A♭	B♭	Cm	D°
Em	Em	F#°	G	Am	Bm	C	D	E	E	F#m	G#m	A	B	C#m	D#°
Fm	Fm	G°	A♭	B♭m	Cm	D♭	E♭	F	F	Gm	Am	B♭	C	Dm	E°
F#m	F#m	G#°	A	Bm	C#m	D	E	F#	F#	G#m	A#m	B	C#	D#m	E#°
Gm	Gm	A°	B♭	Cm	Dm	E♭	F	G	G	A♭m	B♭m	C♭	D♭	E♭m	F°
G#m	G#m	A#°	B	C#m	D#m	E	F#	G#	G#	Am	Bm	C	D	Em	F#°
A♭m	A♭m	B°	C♭	D♭m	E♭m	F♭	G♭	A♭	A♭	B♭m	Cm	D♭	E♭	Fm	G°
Am	Am	B°	C	Dm	Em	F	G	A	A	Bm	C#m	D	E	F#m	G#°
A#m	A#m	B#°	C#	D#m	E#m	F#	G#	A#	A#	Bm	C#m	D	E	F#m	G#°
B♭m	B♭m	C°	D♭	E♭m	Fm	G♭	A♭	B♭	B♭	Cm	Dm	E♭	F	Gm	A°
Bm	Bm	C#°	D	Em	F#m	G	A	B	B	C#m	D#m	E	F#	G#m	A#°

- The similarity of each chord in the individual to the previous two chords in the individual. Where we:
 - Punish the individual for having more than two consecutive chords that are the same.
 - Reward the individual for having exactly two consecutive chords that are the same.

- **Selection technique:**
 - **Selection method:** Tournament.
 - The selection phase selects half of the population to remain, the other half we remove from the population.
 - **Selection process:**
 - We select half of the population based on a random tournament bracket, where random pairs of individuals are put ahead of each other.
 - The Winner gets added to the next population.
 - The Loser is removed fully.
- **Crossover technique:**
 - **Crossover method:** n-point crossover.
 - The crossover phase chooses two random parents from the selected individuals. And they mate and produce two children.
 - **Mating process:**
 - We iterate through the chords of the parents and do the following with a 50% probability:
 - Child1 gets the chord from Parent1, Child2 gets the chord from Parent2.
 - Child1 gets the chord from Parent2, Child2 gets the chord from Parent1.
- **Mutation technique:**
 - **Mutation method:** replacement of a chord with a random new one.
 - The mutation phase takes the new children produced by the crossover phase, mutates them, and then returns the mutated children.
 - **Mutation process:**
 - We iterate through the chords of the child and with a 5% probability:
 - we replace the current chord with a new random chord based on a random root and a random octave

- **Population Size:**

- **Population size = 1024**
- The population size is set before running the evolutionary algorithm, and at the end of each generation we get a new population with the same size.
- The population size I found most suitable is 1024, because:
 - Less than that might lead to getting stuck in local maxima.
 - More than that would make the code very slow.
- The population size needs to be a number that is divisible by two because:
 - The selection phase selects half of the population size
 - The crossover and mutation phases generates a new half of the population
- If the population size is not an even number some errors might occur.

- **Stopping Criteria:**

- Due to the fact that there is no clear definition of “perfect” music, I opted with the maximum number of generations method.
- The maximum number of generation method makes the evolution run for a set number of generations and then terminates.

- **Number of generations:**

- **Maximum number of generations = 5000**
- The number of generations I found most suitable is 5000, because :
 - Less than that would not lead to a relatively good generation.
 - More than that would not lead to a significantly better generation.

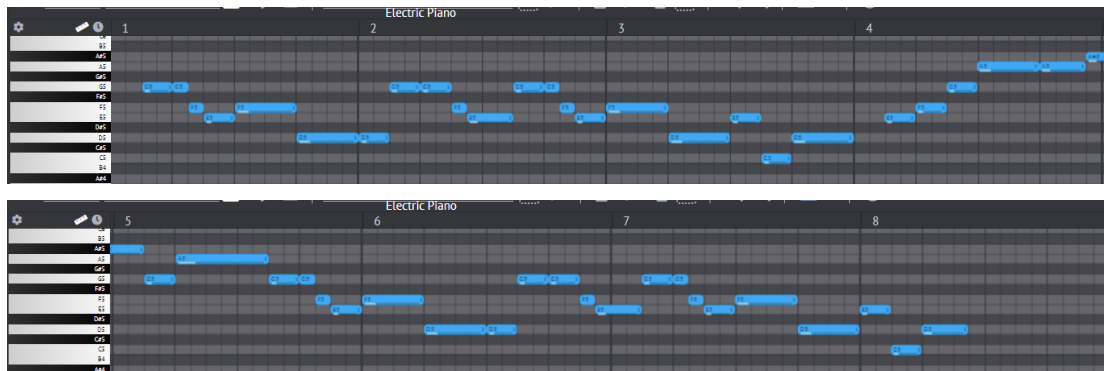
- **Generation Time:**

- The generation time depends on:
 - Population size.
 - Length of individuals.
 - Number of generations.
- On average the generation times for each input is:
 - Input 1:
 - Population size = 1024
 - Length of individual = 29 chord
 - Number of generations = 5000
 - **Generation time = 20 minutes**
 - Input 2:
 - Population size = 1024
 - Length of individual = 32 chord
 - Number of generations = 5000
 - **Generation time = 23 minutes**
 - Input 3:
 - Population size = 1024
 - Length of individual = 44 chord
 - Number of generations = 5000
 - **Generation time = 36 minutes**

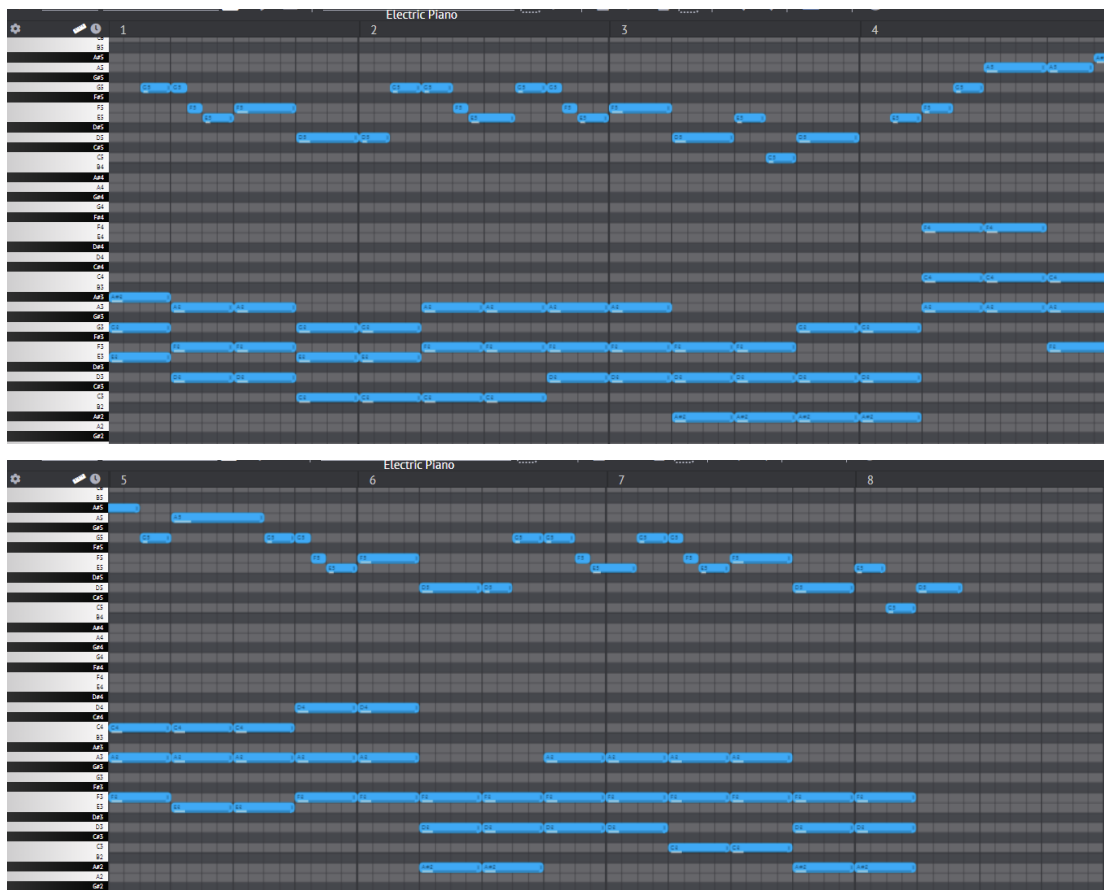
Input - Output:

- **Note:** To listen to the output you can use [OnlineSequencer](#)

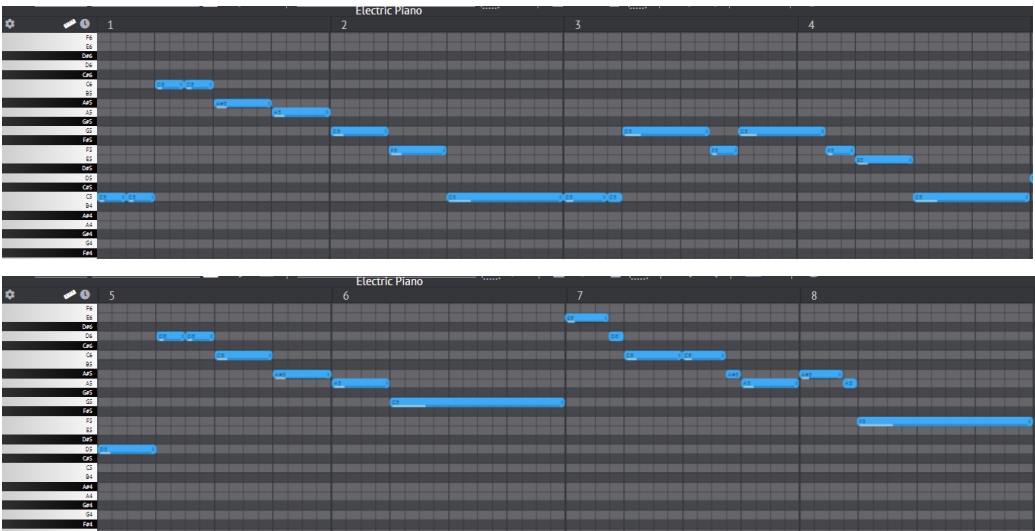
- **Input 1:**



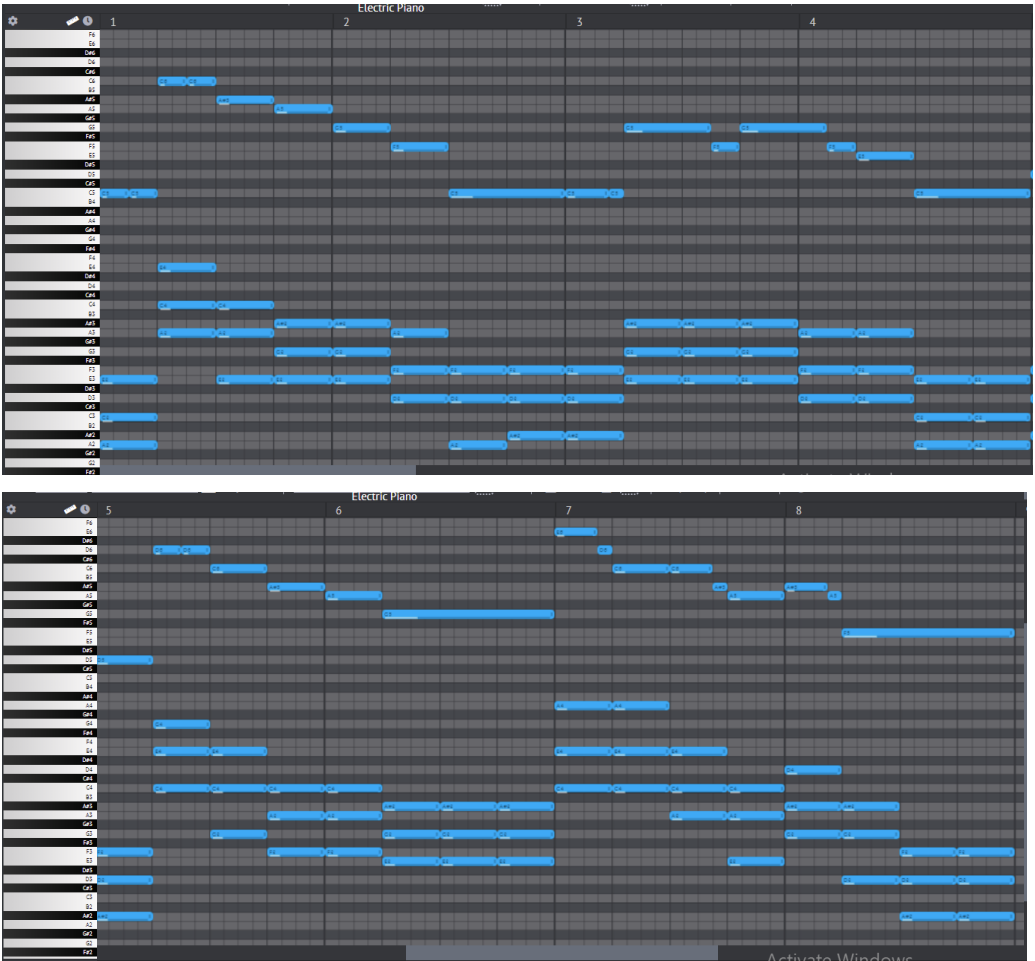
- **Output 1:**



● Input 2:



● Output 2:



● **Input 3:**



● **Output 3:**

