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**Algorithm 1** Marie Curie Genetic Algorithm Overview

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- 1: Let  $NUM\_GENS = 100$  represent the number of generations to evolve
  - 2: Let  $POP\_SIZE = 100$  represent the size of the total population
  - 3: Let  $NBPG = 50$  represent the number of new candidates to breed every generation
  - 4: Let  $\mathbf{S}_{min}$  and  $\mathbf{S}_{max}$  represent the vectors of servo limits
  - 5: Let  $L_R$  represent the landmarks in the reference image
  - 6: Let  $P = \{C_i, i \in \{0, 1, \dots, POP\_SIZE\}\}$  where each  $C_i$  is a candidate expression
  - 7: **for**  $i \in \{0, 1, \dots, POP\_SIZE - 1\}$  **do**
  - 8:     Initialize  $\mathbf{C}_i.\mathbf{chromosome}$  to  $\mathcal{U}(\mathbf{S}_{min}, \mathbf{S}_{max})$
  - 9:     Score each candidate s.t.  $C_i.score = mc\_score()$
  - 10: **end for**
  - 11: **for**  $g \in \{0, 1, \dots, NUM\_GENS\}$  **do**
  - 12:     Let  $N = \{N_i, i \in \{0, 1, \dots, NBPG\}\} = mc\_breed\_new(NBPG)$  be the set of new candidates for this gen
  - 13:     Score  $N$  s.t.  $N_i.score = mc\_score(\mathbf{N}_i.\mathbf{chromosome}, L_R)$
  - 14:     Set  $C = C \cup N$
  - 15:     Remove the  $NBPG$  elements with the worst score from  $C$
  - 16: **end for**
  - 17: **return**  $C_{opt} = C_i$  where  $i = \underset{i}{\operatorname{argmin}}\{C_i.score\}$
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**Algorithm 2** MC Score Algorithm ( $\text{mc\_score}(X, L'_R)$ )

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- 1: Let  $M = 68$  be the number of landmarks recognized in an image (constrained by dlib landmark recognition algorithm)
  - 2: Let  $D = 2$  be the dimensionality of a single landmark
  - 3: Let  $L \in \mathbb{R}^{D,M}$  represent the raw pixel valued landmarks
  - 4: Let  $L'_R$  be the reference landmarks to score against
  - 5: Let  $X$  be the chromosome to score
  - 6: Actuate  $X$  on the face using the Maestro Controller
  - 7: Capture an image,  $I_C$ , of the face
  - 8: Get the bounding box  $B$  of the face
  - 9: Detect the center of a rectangular bounding box,  $B_C = [x, y]^T \in \mathbb{R}^2$  of the face
  - 10: Detect raw landmarks for cand. image  $L_C = \text{dlib\_predictor}(I_C).landmarks$
  - 11: Calculate new "centered" landmarks  $L_C^C = L_C - (B_C \otimes \mathbf{1}^T)$
  - 12: Calculate the horizontal scaling by  $x_s = B_{Right} - B_{C,x}$
  - 13: Calculate the vertical scaling by  $y_s = B_{Bottom} - B_{C,y}$
  - 14: Get normalized landmarks  $L'_C = L_C^C \oslash ([x_s, y_s]^T \otimes \mathbf{1}^T)$
  - 15: Calculate difference in corresponding landmarks between the reference image and the candidate image  $L_D = L'_C - L'_R$
  - 16: **return**  $\|L_D\|_F = \sqrt{\text{tr}(L_D^T L_D)}$  where  $L_D^T L_D \in \mathbb{R}^{(M,M)}$  is the Gramian matrix of  $L_D$  and the diagonal elements of  $L_D^T L_D$  represent the euclidean distance error of each landmark. This can also be expressed as the square root of the sum of the squared errors.
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**Algorithm 3** MC Breed New ( $\text{mc.breed\_new}(N_C, N_{CpP})$ )

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- 1: Let  $N_C$  be the number of new candidates to breed
  - 2: Let  $N_{CpP}$  be the maximum number of candidates to breed for a single set of parents
  - 3: Calculate the number of pairs of parents  $N_{pP}$
  - 4: Get the mating pool  $M$  which is comprised of the best (lowest) scoring  $2N_{pP}$  candidates in  $C$
  - 5: Create  $N_{pP}$  pairs of parents,  $P_p$ , by pairing the highest scoring parents together (i.e.  $P_p = \{\{M_0, M_1\}, \{M_2, M_3\}, \dots\}$ )
  - 6: Initialize  $N = \emptyset$  where  $N$  is the set of all new candidates bred this generation.
  - 7: **for**  $p \in P_p$  **do**
  - 8:     **for**  $j \in \{0, 1, \dots, N_{CpP} - 1\}$  **do**
  - 9:         Let  $C_N$  represent the new candidate
  - 10:         Randomize order of  $p$
  - 11:         Randomly select location  $i$  in which to perform crossover of chromosomes
  - 12:         Set  $\mathbf{C_N.chromosome} = [p_0[:i]^T, p_1[i:]^T]^T$  representing the concatenation of the two parent chromosomes sliced at index  $i$
  - 13:         Set  $\mathbf{C_N.chromosome} = \mathcal{N}(\mathbf{C_N.chromosome}, 50^2)$  representing a mutation of the inherited chromosome with each gene varying in a normal distribution parameterized by  $\mu = \mathbf{C_N.chromosome}$  and  $\sigma^2 = 50^2$  where 50 is the standard deviation in quarter microseconds
  - 14:         Saturate values in  $\mathbf{C_N.chromosome}$  to safe limits for servos
  - 15:         Append  $C_N$  to  $N$  by  $N = N \cup C_N$
  - 16:     **end for**
  - 17: **end for**
  - 18: **return**  $N$
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