# Comparing BLP and a Genetic Algorithm

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The arm being optimized,

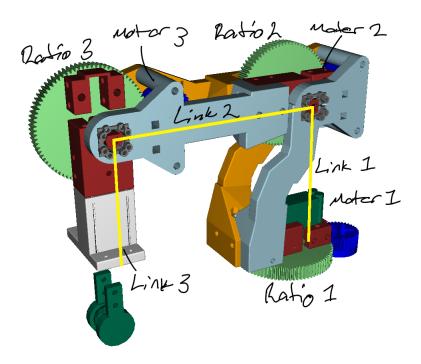


Figure 1: RBE 3001 3DOF Arm

## 1 BLP

#### 1.1 Feature Matrix

The feature matrix of a motor module is given by

$$F_{m} = \begin{bmatrix} \frac{\tau^{(1)}}{G^{(1)}} & \frac{\tau^{(N)}}{G^{(N)}} \\ \omega^{(1)}G^{(1)} & \omega^{(N)}G^{(N)} \\ P^{(1)} & P^{(N)} \\ M^{(1)} & M^{(N)} \\ G^{(1)} & G^{(N)} \\ \ln(\omega^{(1)}G^{(1)}) & \ln(\omega^{(N)}G^{(N)}) \\ R_{1} & R_{1} \\ R_{2} & \cdots & R_{2} \\ R_{3} & R_{3} \\ \ln(R_{1}) & \ln(R_{1}) \\ \ln(R_{2}) & \ln(R_{2}) \\ \ln(R_{3}) & \ln(R_{2}) \\ \ln(R_{3}) & \ln(R_{1} + R_{2} + R_{3}) \\ \ln(R_{1} + R_{2} + R_{3}) & \ln(R_{1} + R_{2} + R_{3}) \\ \ln(R_{2} + R_{3}) & \ln(R_{2} + R_{3}) \\ M^{(1)}R_{1} & M^{(N)}R_{1} \\ M^{(1)}R_{2} & T_{c}^{(2)} \\ \tau^{(2)} & \tau^{(2)}_{c} \\ \tau^{(3)} & T_{c}^{(3)} \end{bmatrix}$$

where  $\tau^{(i)}$  is the stall torque in Newton-meters for motor i,  $\omega^{(i)}$  is the free speed in radians per second for motor i,  $P^{(i)}$  is the price of motor i in USD,  $M^{(i)}$  is the mass in kilograms of motor i,  $G^{(i)}$  is the gear ratio on motor i, and  $R_i$  is the r parameter of link i.  $\tau^{(i)}_c$  is the required joint torque at the target  $(T_x, 0, T_z)$  for joint i given by

$$\tau_c^{(i)} = \left(J(\boldsymbol{\theta})^\top \begin{bmatrix} 0\\0\\F_c \end{bmatrix}\right)_{(i,1)} \tag{2}$$

where  $J(\theta)$  is the manipulator jacobian at configuration  $\theta$  ( $\theta$  is computed using an inverse kinematics algorithm parameterized over the link lengths with target position  $(T_x, 0, T_z)$  supplied from the configuration file).

#### 1.2 Variables

 $\tau_j$  denotes the  $\tau$  of slot j,  $\omega_j$  denotes the  $\omega$  of slot j, and  $R_j$  denotes the R of slot j.  $\tau_j$  and  $\omega_j$  are implemented using a binary vector for each slot j. Each  $R_j$  is integral and bounded by a minimum and maximum length parameter given in the constraints file.

#### 1.3 Constraints

V is the tip velocity, F is the tip force. The arm is mounted  $90\deg$  off vertical. In places where a quadratic term is in the feature matrix, the implementation pulls the term from the feature matrix instead of introducing a quadratic constraint (e.x.,  $M_2R_1$  is taken from the feature matrix).

$$\tau_1 \ge F(R_1 + R_2 + R_3)$$
(3)

$$\tau_2 \ge F(R_2 + R_3) + M_3 G R_2 \tag{4}$$

$$\tau_3 \ge FR_3 \tag{5}$$

$$\ln(\omega_1) \ge \ln(V) - \ln(R_1 + R_2 + R_3) \tag{6}$$

$$ln(\omega_2) \ge ln(V) - ln(R_2 + R_3)$$
(7)

$$ln(\omega_3) \ge ln(V) - ln(R_3)$$
(8)

$$R_1 + R_2 + R_3 = 0.4 (9)$$

$$\tau_2 \ge \tau_c^{(2)} \tag{10}$$

$$\tau_3 \ge \tau_c^{(3)} \tag{11}$$

## 1.4 Results

BLP took 3.27 seconds to solve and found:

Optimal objective: 42.93

Optimal motors:

Motor("stepperMotor-Pololu35x26", 0.098, 139.626, 12.95, 0.12), ratio=0.14285714285714285,

# 2 Genetic Algorithm

The genetic algorithm was implemented using a constraint handling technique from [1].

# 2.1 Entity

The entity is given by

$$\begin{bmatrix} M_1 & M_2 & M_3 & R_1 & R_2 & R_3 & G_1 & G_2 & G_3 \end{bmatrix}$$
 (12)

where  $M_i$  is the integer index of motor i in the array of motors parsed from the motor options file,  $R_i$  is the real r parameter of link i, bounded by the minimum and maximum length parameter given in the constraints file, and  $G_i$  is the real gear ratio for motor i.

#### 2.2 Constraints

V is the tip velocity, F is the tip force,  $\tau_j$  is the  $\tau$  for the motor in index j, and  $\omega_j$  is the  $\omega$  for the motor in index j. The arm is mounted  $90 \deg$  off vertical.

$$F(R_1 + R_2 + R_3) - \frac{\tau_1}{G_1} \le 0 \tag{13}$$

$$F(R_2 + R_3) + M_3 G R_2 - \frac{\tau_2}{G_2} \le 0 \tag{14}$$

$$FR_3 - \frac{\tau_3}{G_3} \le 0 \tag{15}$$

$$\frac{V}{R_1 + R_2 + R_3} - \omega_1 G_1 \le 0 \tag{16}$$

$$\frac{V}{R_2 + R_3} - \omega_2 G_2 \le 0 \tag{17}$$

$$\frac{V}{R_3} - \omega_3 G_3 \le 0 \tag{18}$$

$$|R_1 + R_2 + R_3 - 0.4| - 0.0001 \le 0 (19)$$

$$\left(\tau_c^{(2)} - \frac{\tau_2}{G_2}\right) + \left(\tau_c^{(3)} - \frac{\tau_3}{G_3}\right) \le 0 \tag{20}$$

 $\tau_c^{(i)}$  is the required joint torque at the target  $(T_x,0,T_z)$  for joint i given by

$$\tau_c^{(i)} = \left(J(\boldsymbol{\theta})^\top \begin{bmatrix} 0\\0\\F_c \end{bmatrix}\right)_{(i,1)} \tag{21}$$

where  $J(\theta)$  is the manipulator jacobian at configuration  $\theta$  ( $\theta$  is computed using an inverse kinematics algorithm parameterized over the link lengths with target position  $(T_x, 0, T_z)$  supplied from the configuration file).

#### 2.3 Fitness

The fitness is the negative of the sum of the price of each motor,

$$-\sum_{i=1}^{N} P_i \tag{22}$$

# 2.4 Crossover

Crossover is performed by taking the motors and gear ratios from one entity and the link lengths from the other entity.

#### 2.5 Mutation

Mutation is performed by selecting a random motor index, a random link length between the minimum and maximum length parameter given in the constraints file, and a random gear ratio. A random gear ratio is selected by first selecting the range of the ratio (50% > 1, 50% < 1) and then sampling a ratio from a uniform distribution of ratios in that range.

# 2.6 Stopping Criteria

The genetic algorithm is stopped after a maximum generation number is reached.

#### 2.7 Results

With a mutation rate of 5%, an elitism ratio of 25%, and a mutation ratio of 5%, and a maximum of 200,000 generations, the genetic algorithm took 95.83 seconds to solve and found: Motor 1=Motor("vexMotor-393", 1.637, 10.471, 14.99, 0.0945)
Motor 2=Motor("vexMotor-393", 1.637, 10.471, 14.99, 0.0945)
Motor 3=Motor("stepperMotor-Pololu35x26", 0.098, 139.626, 12.95, 0.12)
Link 1=100.0
Link 2=170.0
Link 3=130.0
Gear ratio 1=0.33206607726344173
Gear ratio 2=0.8486810156880156
Gear ratio 3=0.1529117565209915
Fitness=-42.93
Constraint values=[-2.9684137354115947, -0.2578238182300667, -0.0034609136279283303, -0.9770638950254984, -5.553205581935878, -13.658149223692268, -0.0001, -7.069767991857092]
Feasible=true

The algorithm was run 20 times. The best fitness per run was,

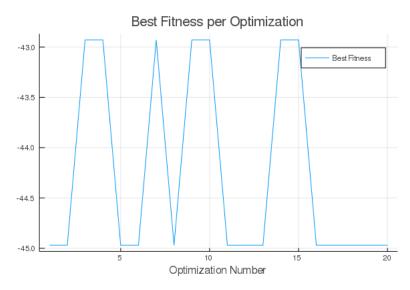


Figure 2: Best Fitness per Optimization Run

# 3 Summary

All algorithms found an optimal solution to the problem. BLP the fastest at 3.27 seconds and the genetic algorithm was the slowest at 95.83 seconds.

## References

[1] Adam Chehouri, Rafic Younes, Jean Perron, and Adrian Ilinca. A constraint-handling technique for genetic algorithms using a violation factor. arXiv preprint arXiv:1610.00976, 2016.