

Model assisted (1+1)ES

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1 Algorithms

1.1 Algorithm 1

Normal (1+1) ES

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1:  $N \leftarrow 400$ 
2: function FIT( $x$ )
3:    $y \leftarrow \sum_{i < N} (x_i - 1)^2$ 
4:   return( $y$ )
5: end function
6: function DIST( $x$ )
7:    $y \leftarrow \sqrt{\sum_{i < N} (x_i - 1)^2}$ 
8:   return( $y$ )
9: end function
10: Initialize( $ind, \sigma^*, prate^* \leftarrow 0$ )
11: for  $i \leq MAXGen$  do
12:    $i++$ 
13:    $\sigma \leftarrow \frac{\sigma^* * ||ind - y||}{N}$ 
14:    $ind2 \leftarrow ind + \sigma * N(0, I)$ 
15:    $newfit \leftarrow FIT(ind2)$ 
16:   if  $newfit \leq bestfit$  then
17:      $prate^* \leftarrow prate^* - N * \log \frac{DIST(ind2)}{DIST(ind)}$ 
18:      $ind \leftarrow ind2$ 
19:      $bestfit \leftarrow newfit$ 
20:   end if
21: end for
22:  $prate^* \leftarrow prate^* / i$ 
```

1.2 Algorithm 2

Model Assisted (1+1)ES

```

1:  $N \leftarrow 400$ 
2: function FIT( $x$ )
3:    $y \leftarrow \sum_{i < N} (x_i - 1)^2$ 
4:   return( $y$ )
5: end function
6: function DIST( $x$ )
7:    $y \leftarrow \sqrt{\sum_{i < N} (x_i - 1)^2}$ 
8:   return( $y$ )
9: end function
10: initialize( $ind, \sigma^*, \sigma_e^*, prate^* \leftarrow 0$ )
11: for  $i \leq MAXGen$  do
12:    $i++$ 
13:    $ind2 \leftarrow ind$ 
14:   while  $j \leq MAXModel$  do
15:      $j++$ 
16:      $\sigma \leftarrow \frac{\sigma^* * Dist(ind)}{N}$ 
17:      $ind3 \leftarrow ind2 + \sigma * N(0, I)$ 
18:      $\sigma_e \leftarrow \frac{2 * \sigma_e^* * Dist(ind)^2}{N}$ 
19:      $fit3 \leftarrow FIT(ind2) + \sigma_e * N(0, 1)$ 
20:     if  $fit3 \leq bestfit$  then
21:        $ind2 \leftarrow ind3$ 
22:       Break
23:     end if
24:   end while
25:    $newfit = FIT(ind2)$ 
26:   if  $newfit \leq bestfit$  then
27:      $prate^* \leftarrow prate^* - N * \log \frac{DIST(ind2)}{DIST(ind)}$ 
28:      $ind \leftarrow ind2$ 
29:      $bestfit \leftarrow newfit$ 
30:   end if
31: end for
32:  $prate^* \leftarrow prate^* / i$ 

```

1.3 Algorithm 3

Model Assisted (1+1)ES without model generation limit

```

1:  $N \leftarrow 400$ 
2: function FIT( $x$ )
3:    $y \leftarrow \sum_{i < N} (x_i - 1)^2$ 
4:   return( $y$ )
5: end function
6: function DIST( $x$ )
7:    $y \leftarrow \sqrt{\sum_{i < N} (x_i - 1)^2}$ 
8:   return( $y$ )
9: end function
10: initialize( $ind, \sigma^*, \sigma_e^*, prate^* \leftarrow 0$ )
11: for  $i \leq MAXGen$  do
12:    $i++$ 
13:    $ind2 \leftarrow ind$ 
14:   while true do
15:      $j++$ 
16:      $\sigma \leftarrow \frac{\sigma^* * Dist(ind)}{N}$ 
17:      $ind3 \leftarrow ind2 + \sigma * N(0, I)$ 
18:      $\sigma_e \leftarrow \frac{2 * \sigma_e^* * Dist(ind)^2}{N}$ 
19:      $fit3 \leftarrow FIT(ind2) + \sigma_e * N(0, 1)$ 
20:     if  $fit3 \leq bestfit$  then
21:        $ind2 \leftarrow ind3$ 
22:       Break
23:     end if
24:   end while
25:    $newfit = FIT(ind2)$ 
26:   if  $newfit \leq bestfit$  then
27:      $prate^* \leftarrow prate^* - N * \log \frac{DIST(ind2)}{DIST(ind)}$ 
28:      $ind \leftarrow ind2$ 
29:      $bestfit \leftarrow newfit$ 
30:   end if
31: end for
32:  $prate^* \leftarrow prate^* / i$ 

```

1.4 Algorithm 4

Step-size Adaptive Model Assisted (1+1)ES

```

1:  $N \leftarrow 400$ 
2: function FIT( $x$ )
3:    $y \leftarrow \sum_{i < N} (x_i - 1)^2$ 
4:   return( $y$ )
5: end function
6: function DIST( $x$ )
7:    $y \leftarrow \sqrt{\sum_{i < N} (x_i - 1)^2}$ 
8:   return( $y$ )
9: end function
10: initialize( $ind, \sigma_e^*, prate^* \leftarrow 0$ )
11: for  $i \leq MAXGen$  do
12:    $i++$ 
13:    $ind2 \leftarrow ind$ 
14:    $flag \leftarrow 0$ 
15:   for  $j \leq MAXModel$  do
16:      $j++$ 
17:      $ind3 \leftarrow ind2 + \sigma * N(0, I)$ 
18:      $\sigma_e \leftarrow \frac{2 * \sigma_e^* * Dist(ind)^2}{N}$ 
19:      $fit3 \leftarrow FIT(ind2) + \sigma_e * N(0, 1)$ 
20:     if  $fit3 \leq bestfit$  then
21:        $flag \leftarrow 1$ 
22:        $ind2 \leftarrow ind3$ 
23:       Break
24:     end if
25:   end for
26:    $newfit = FIT(ind2)$ 
27:    $\sigma \leftarrow \sigma * \exp^{\frac{1}{N}}(flag - \alpha)$ 
28:   if  $newfit \leq bestfit$  then
29:      $prate^* \leftarrow prate^* - N * \log \frac{DIST(ind2)}{DIST(ind)}$ 
30:      $ind \leftarrow ind2$ 
31:      $bestfit \leftarrow newfit$ 
32:   end if
33: end for
34:  $prate^* \leftarrow prate^* / i$ 

```

2 Results

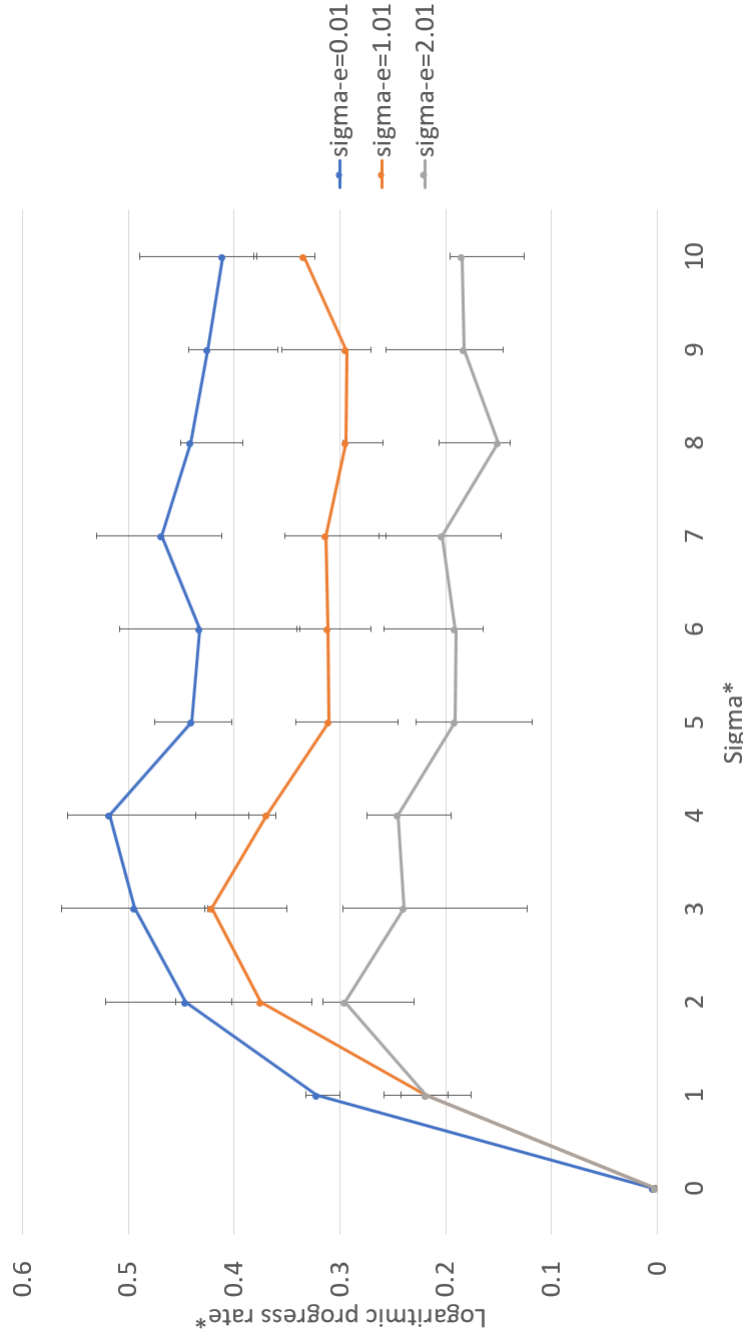


Figure 1: Normalized logarithmic progress rate as a function of normalized mutation strength. Algorithm 3 - Model assisted (1+1)ES without model generation limit, 4 Dimensions, $Y = (X - 1)^2$. Each point represents the median result of 5 trails, error bars show the range of results for each point. (5 trails) (100 original fitness generations)

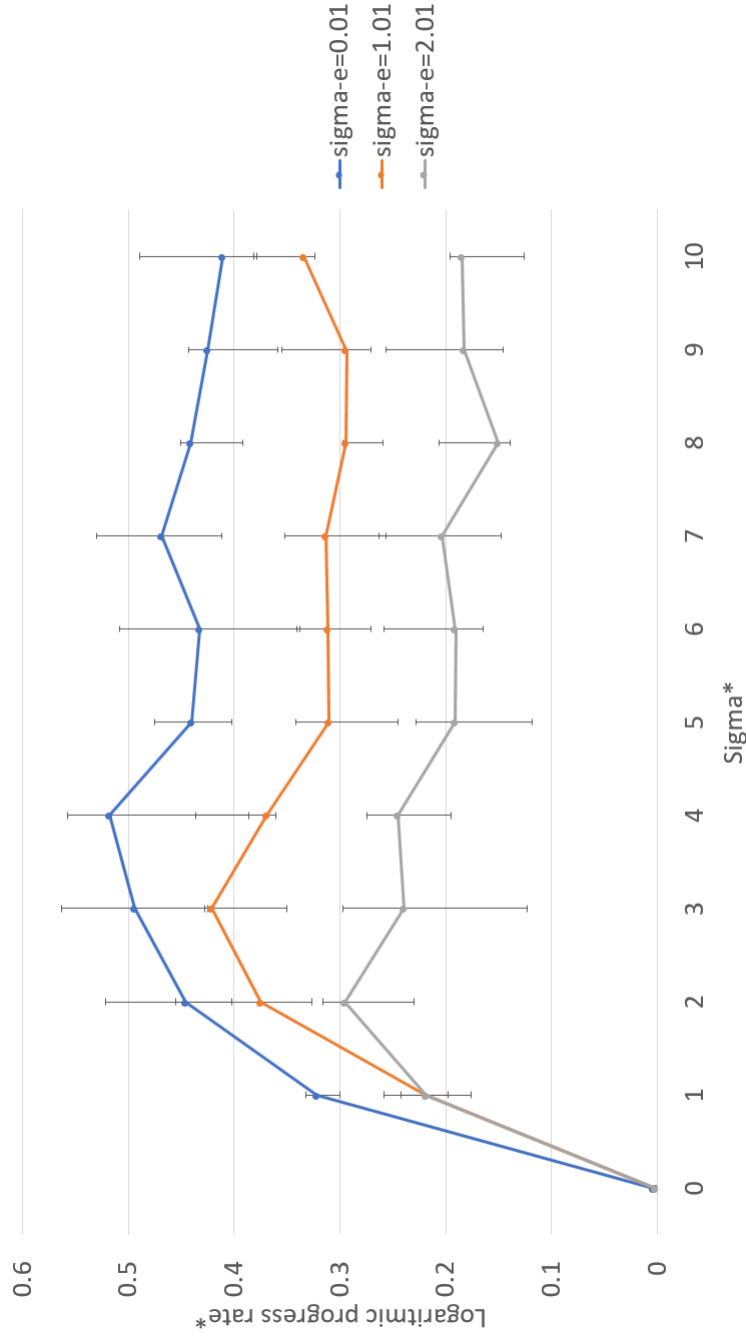


Figure 2: Normalized logarithmic progress rate as a function of normalized mutation strength. Algorithm 3 - Model assisted (1+1)ES without model generation limit, 40 Dimensions, $Y = (X - 1)^2$. Each point represents the median result of 5 trails, error bars show the range of results for each point. (5 trails) (900 original fitness generations)

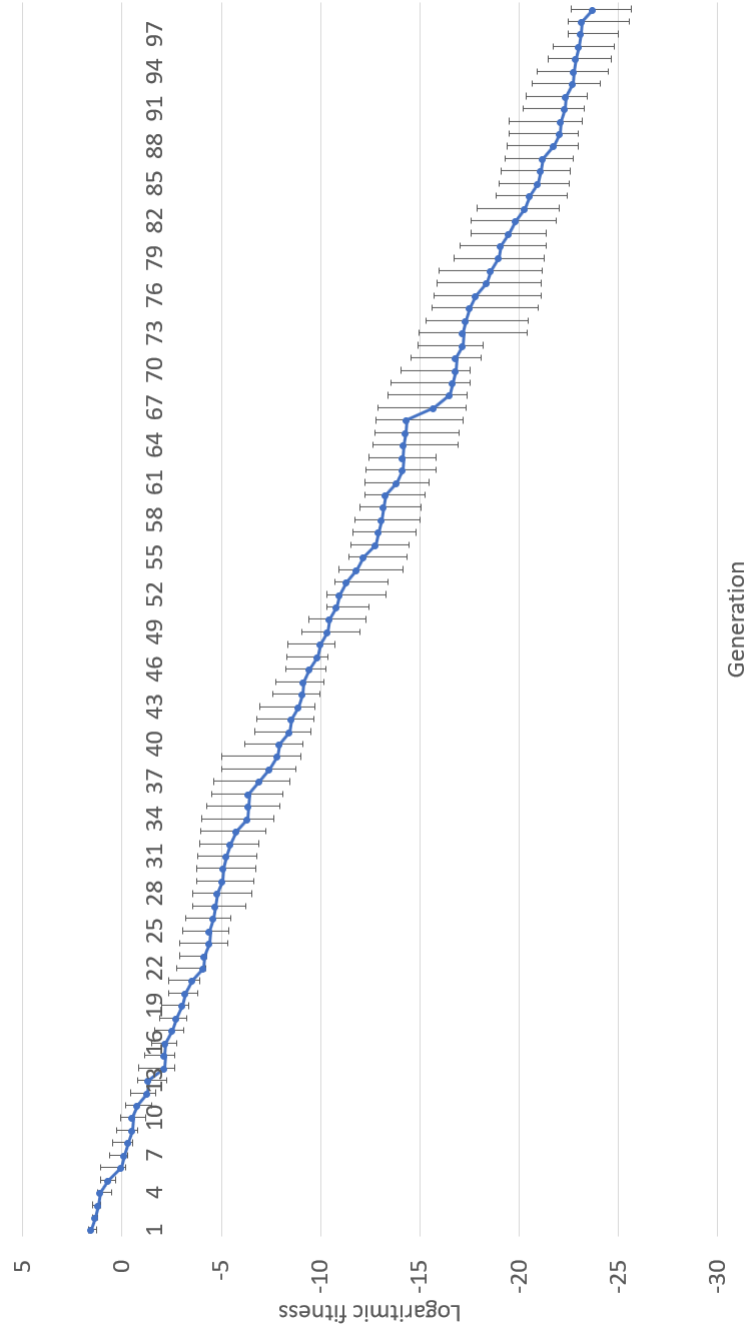


Figure 3: Normalized fitness as a function of normalized mutation strength. Algorithm 3- Model assisted (1+1)ES without model generation limit, 4 Dimensions, $Y = (X - 1)^2$. Each point represents the median result of 5 trails, error bars show the range of results for each point.(5 trails) (100 original fitness generations)

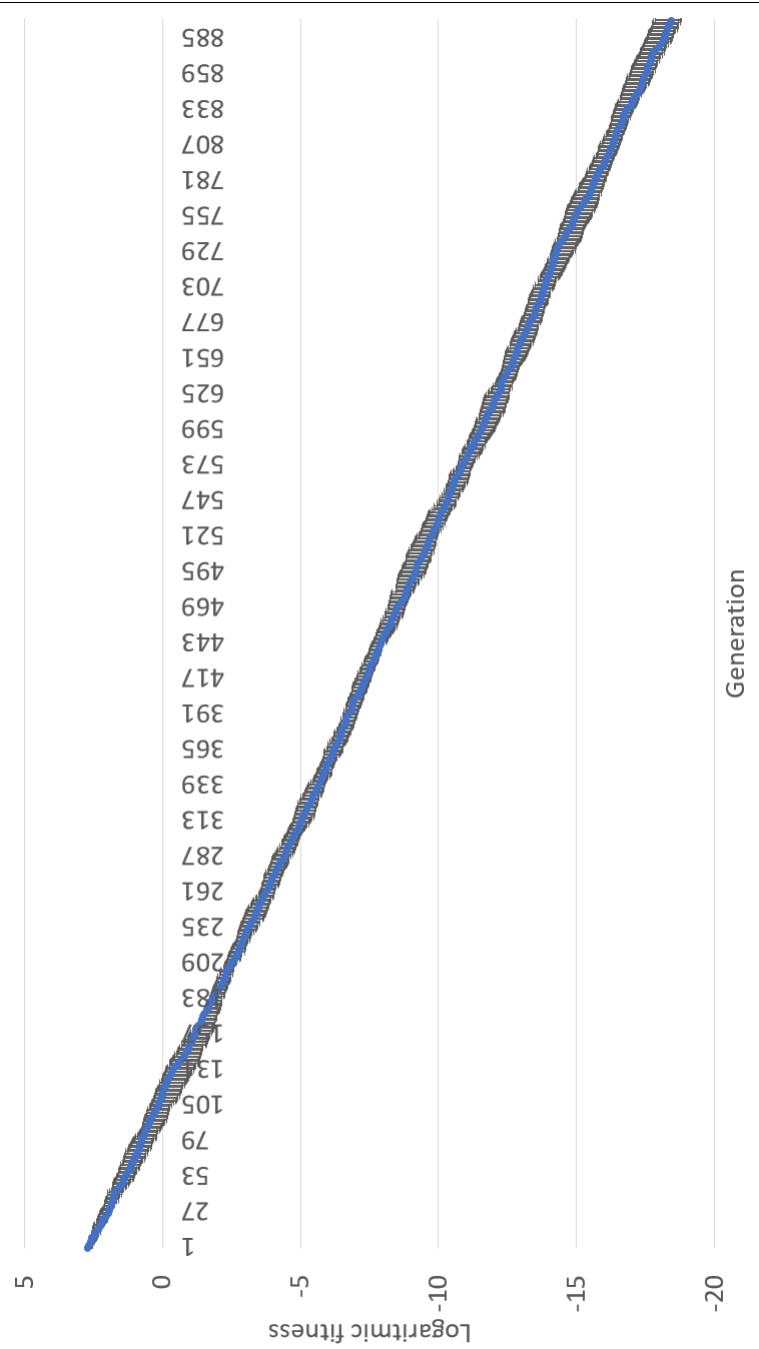


Figure 4: Normalized fitness as a function of normalized mutation strength. Algorithm 3- Model assisted (1+1)ES without model generation limit, 40 Dimensions, $Y = (X - 1)^2$. Each point represents the median result of 5 trails, error bars show the range of results for each point. (5 trails) (900 original fitness generations)

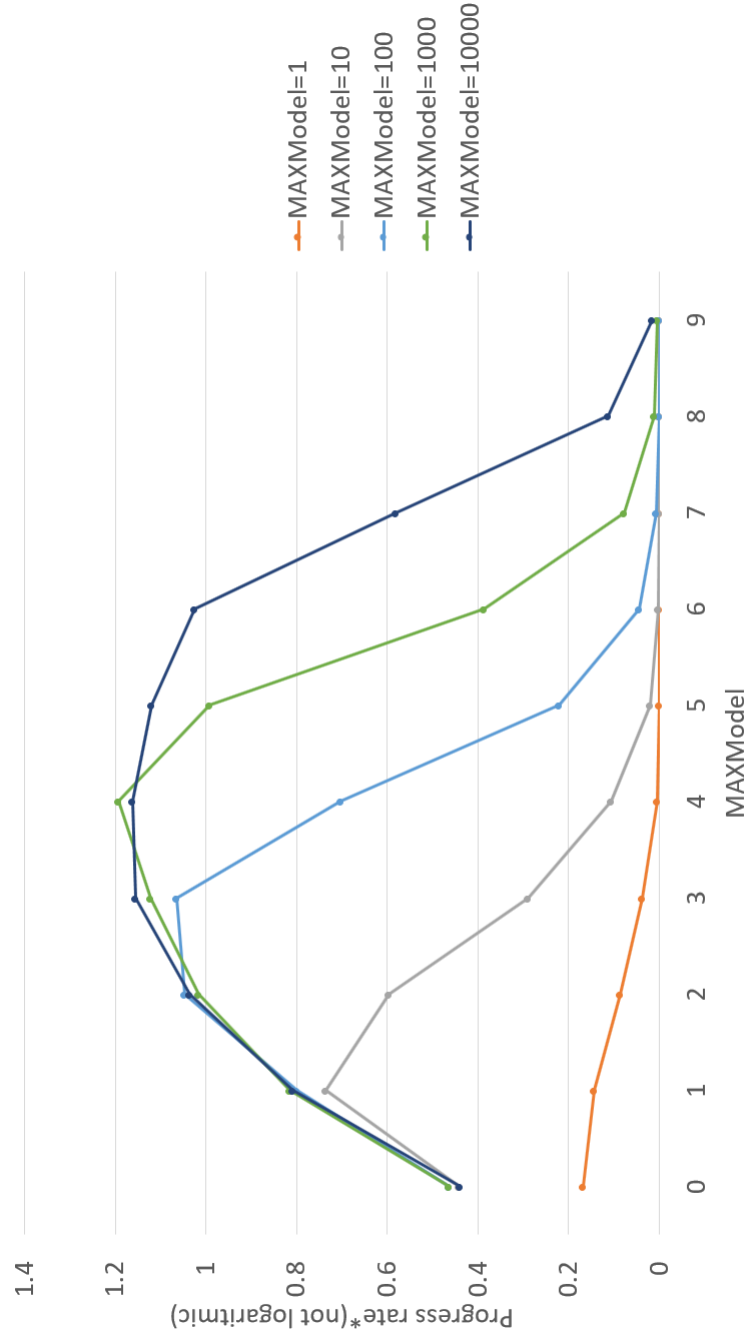


Figure 5: Normalized progress rate as a function of maximum number of model generations. Algorithm 2-Model assisted (1+1)ES, 40 Dimensions, $Y = (X - 1)^2$. Each point represents the median result of 5 trails, error bars show the range of results for each point. (5 trails) (900 original fitness generations)

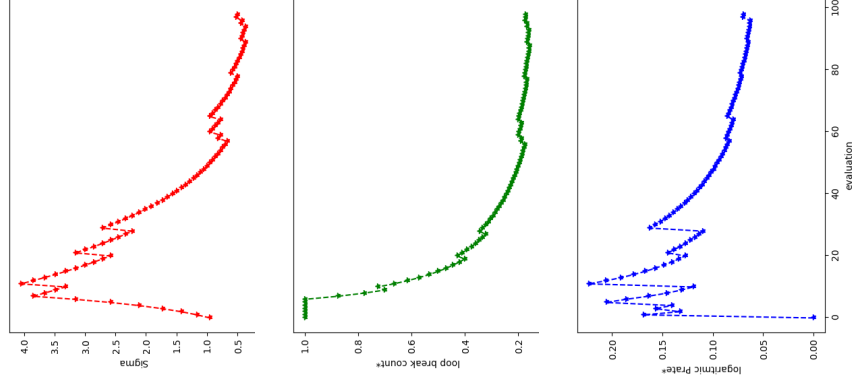


Figure 6: Sigma, logarithmic progress rate and number of generations that in less than 1000 model generations the algorithm finds a better model and exits the modeling loop in proportion to the number of original fitness generations. Model assisted (1+1)ES, 4 Dimensions, $Y = (X - 1)^2$. ($\sigma_{initialvalue}^* = 1$)($\alpha = 0.2$)(100 original fitness generations)(1000 model-generation)

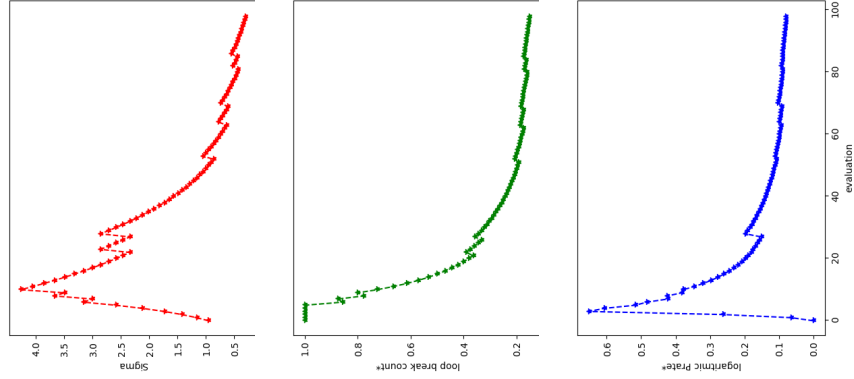


Figure 7: Sigma, logarithmic progress rate and number of generations that in less than 10000 model generations the algorithm finds a better model and exits the modeling loop in proportion to the number of original fitness generations. Model assisted (1+1)ES, 4 Dimensions, $Y = (X - 1)^2$. ($\sigma_{initialvalue}^* = 1$)($\alpha = 0.2$)(100 original fitness generations)(10000 model-generation)

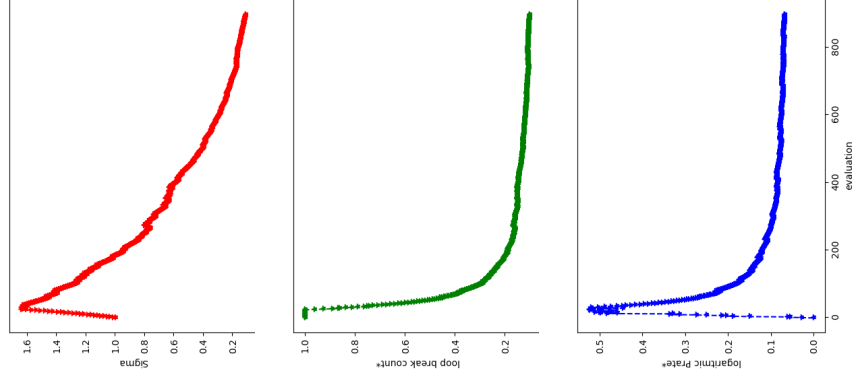


Figure 8: Sigma, logarithmic progress rate and number of generations that in less than 1000 model generations the algorithm finds a better model and exits the modeling loop in proportion to the number of original fitness generations. Model assisted (1+1)ES, 40 Dimensions, $Y = (X-1)^2$. ($\sigma_{initialvalue}^* = 1$)($\alpha = 0.2$)(100 original fitness generations)(1000 model-generation)

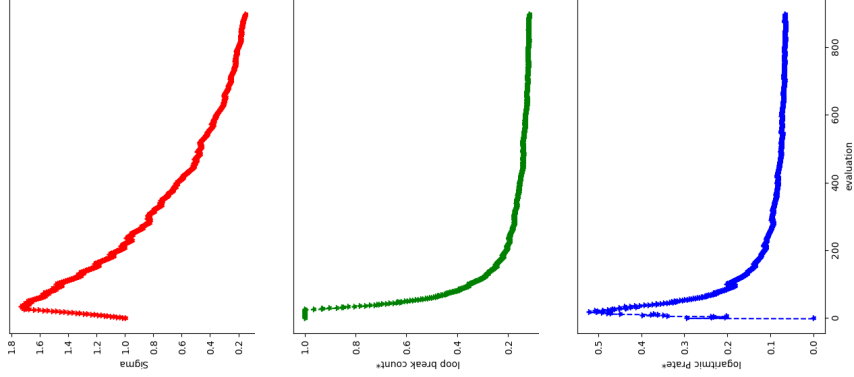


Figure 9: Sigma, logarithmic progress rate and number of generations that in less than 10000 model generations the algorithm finds a better model and exits the modeling loop in proportion to the number of original fitness generations. Model assisted (1+1)ES, 40 Dimensions, $Y = (X-1)^2$. ($\sigma_{initialvalue}^* = 1$)($\alpha = 0.2$)(100 original fitness generations)(10000 model-generation)