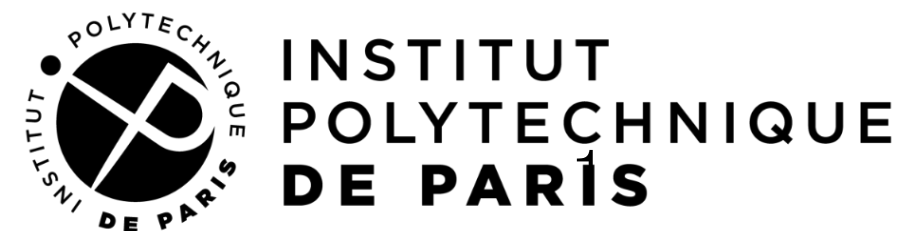


It's Always the Step-Size

A report of a semi-practitioner

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Overview

- empirical research (my methodology)
- what is cGAs K in CMA-ES?
 K is considered to be a replacement/analog for population size, $1/K$ is a learning rate
- in Evolution Strategies (ES)
 - the optimal standard deviation for sampling is proportional to the parent number μ
 - diversity is close to divergence
 - premature diversity loss due to genetic drift happens in noisy landscapes
 - it's always the step-size

The CMA-ES is an EDA

Input: $\mathbf{m} \in \mathbb{R}^n$; $\sigma \in \mathbb{R}_+$; $\lambda \in \mathbb{N}_{\geq 2}$, usually $\lambda \geq 5$, default $4 + \lfloor 3 \log n \rfloor$

Initialize $\mathbf{C} = \mathbf{I}$

While not *terminate*

$\mathbf{x}_i = \mathbf{m} + \sigma \mathbf{y}_i$, where $\mathbf{y}_i \sim \mathcal{N}_i(\mathbf{0}, \mathbf{C})$ for $i = 1, \dots, \lambda$ sampling

$\mathbf{m} \leftarrow \mathbf{m} + c_m \sigma \mathbf{y}_w$, where $\mathbf{y}_w = \sum_{i=1}^{\mu} w_{\text{rk}(i)} \mathbf{y}_i$ update mean

$\mathbf{C} \leftarrow \mathbf{C} + c_\mu \sum_{i=1}^{\lambda} w_{\text{rk}(i)} (\mathbf{y}_i \mathbf{y}_i^T - \mathbf{C})$ update $\mathbf{C}$

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Initialize $\mathbf{C} = \mathbf{I}$, and $\mathbf{p}_c = \mathbf{0}$, $\mathbf{p}_\sigma = \mathbf{0}$

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$\mathbf{p}_c \leftarrow (1 - c_c) \mathbf{p}_c + \mathbb{1}_{[0, 2n]} \left\{ \|\mathbf{p}_\sigma\|^2 \right\} \sqrt{1 - (1 - c_c)^2} \sqrt{\mu_w} \mathbf{y}_w$ path for $\mathbf{C}$

$\sigma \leftarrow \sigma \times \exp \left(\frac{c_\sigma}{d_\sigma} \left(\frac{\|\mathbf{p}_\sigma\|}{\mathbb{E} \|\mathcal{N}(\mathbf{0}, \mathbf{I})\|} - 1 \right) \right)$ update of σ

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corrections for: positive definiteness guaranty, $\mathbf{p}_c$ variance loss, c_σ and d_σ for large λ

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c_m, c_μ, c_1 are analog
to $1/K$ in cGA,
 d_σ is analog to K

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```

1 %matplotlib widget
2 import cma
3
4 x, es = cma.fmin2(cma.ff.rosen, 14 * [1], 1, {'CMA_diagonal_decoding': True, 'ftarget': 1e-7})
5 es.plot();

```

✓ 2.2s

Python

(5_w,11)-aCMA-ES (mu_w=3.4,w_1=42%) in dimension 14 (seed=553091, Mon Feb 24 13:19:30 2025)

Iterat #Fevals function value axis ratio sigma min&max std t[m:s]

1	11	1.040197601953083e+03	1.0e+00	8.93e-01	7e-01	9e-01	0:00.0
2	22	1.476140839356284e+03	1.1e+00	8.01e-01	6e-01	8e-01	0:00.0
3	33	1.027303494859089e+03	1.1e+00	7.42e-01	6e-01	8e-01	0:00.0
100	1100	1.331908726689222e-01	3.7e+00	3.99e-02	4e-03	1e-02	0:00.1
200	2200	7.625477142936609e-04	2.4e+01	7.53e-03	3e-04	8e-03	0:00.2
265	2915	9.436810372489095e-08	3.0e+01	1.86e-04	6e-06	2e-04	0:00.2

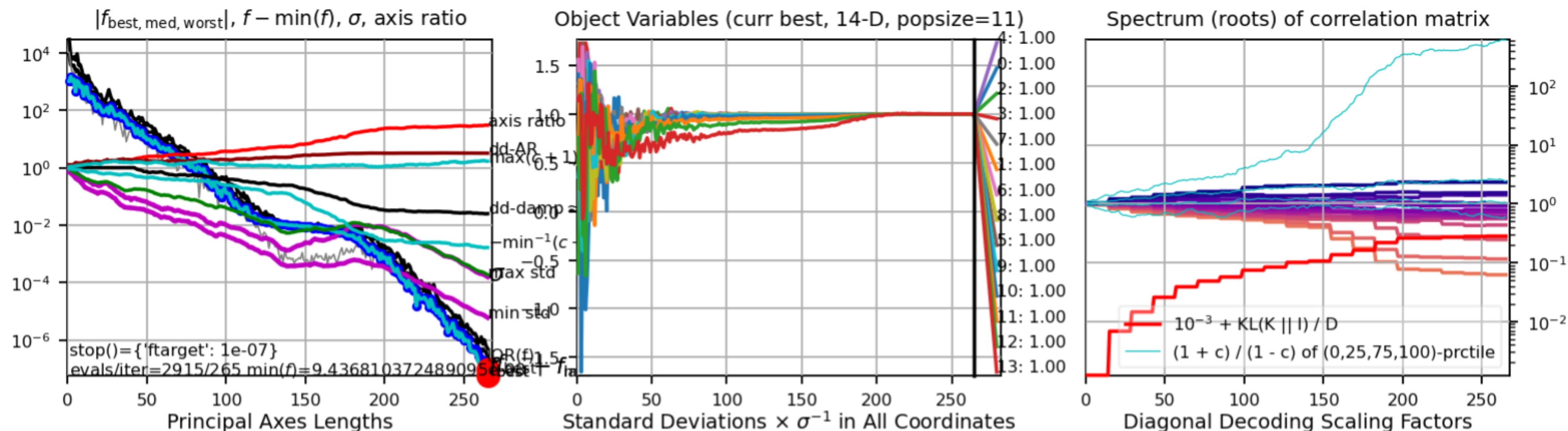
termination on ftarget=1e-07 (Mon Feb 24 13:19:31 2025)

final/bestever f-value = 3.152852e-08 3.152852e-08 after 2916/2916 evaluations

incumbent solution: [1.00000409 1.00000359 1.0000048 1.00000458 1.00000313 1.00000012
0.99999997 1.00000003 ...]

std deviations: [6.00652667e-06 6.04096130e-06 6.75724117e-06 6.71399653e-06
6.68393736e-06 6.42605509e-06 6.86269378e-06 6.56077278e-06 ...]

Figure 325



```

35 1.027503494859089e+03 1.1e+00 7.142e-01 8e-01 8e-01 0:00.0
100 1100 1.331908726689222e-01 3.7e+00 3.99e-02 4e-03 1e-02 0:00.1
200 2200 7.625477142936609e-04 2.4e+01 7.53e-03 3e-04 8e-03 0:00.2
265 2915 9.436810372489095e-08 3.0e+01 1.86e-04 6e-06 2e-04 0:00.2
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```

Figure 325

