

Evaluation of the impact of various modifications to CMA-ES that facilitate its theoretical analysis

BBOB Workshop - GECCO '23 in Lisbon, Portugal

Armand Gissler

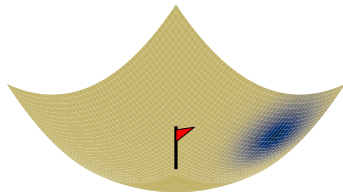
Sunday 16th July, 2023

CMA, École polytechnique & Inria

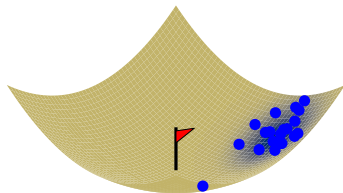


Inria

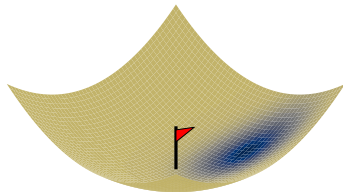
CMA-ES approximates the minimum by a normal distribution



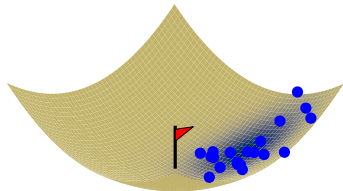
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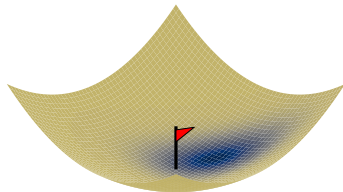
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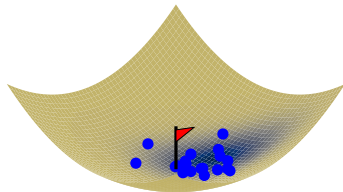
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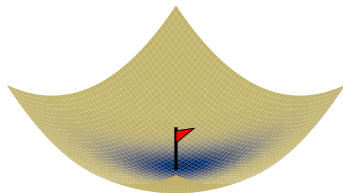
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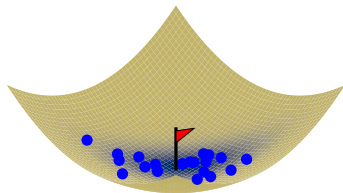
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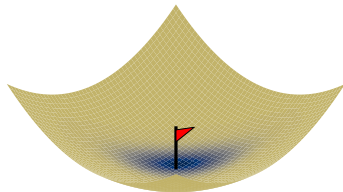
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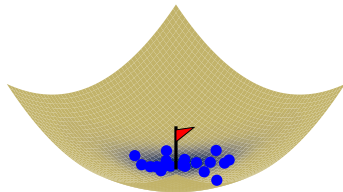
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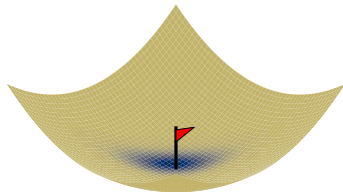
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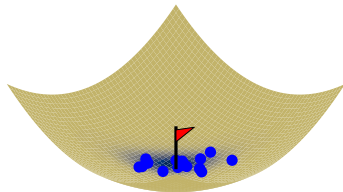
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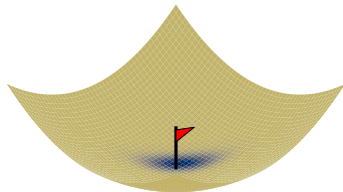
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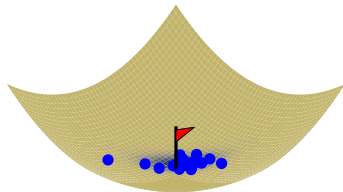
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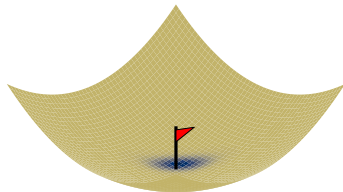
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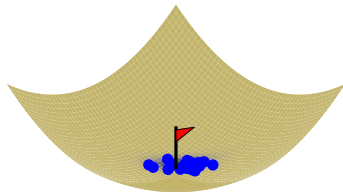
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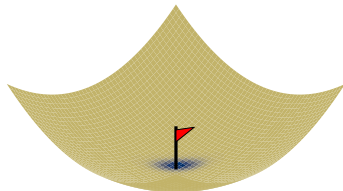
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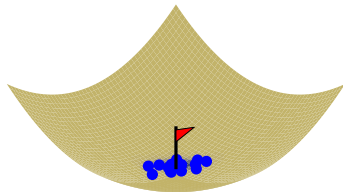
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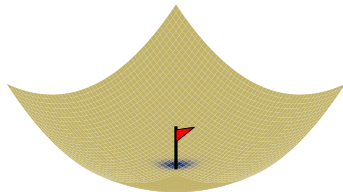
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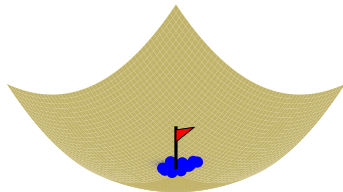
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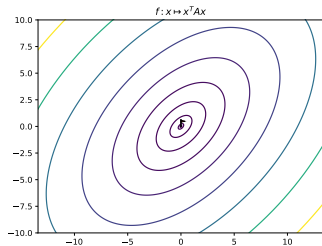
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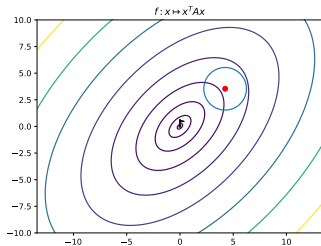
Algorithm 1 default-CMA-ES



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Given: $m_0 \in \mathbb{R}^d$, $\sigma_0 > 0$, $C_0 \in \mathcal{S}_{++}^d$, $p_0^\sigma \in \mathbb{R}^d$, $p_0^c \in \mathbb{R}^d$

For $t = 0, 1, 2, \dots$:

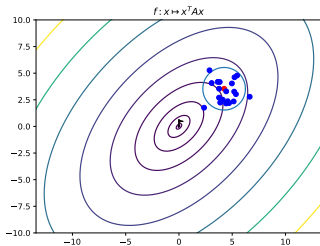


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For $t = 0, 1, 2, \dots$:

- $U_{t+1}^1, \dots, U_{t+1}^\lambda \sim \mathcal{N}(0, I_d)$

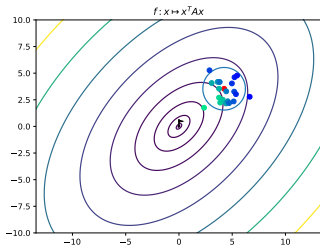


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- $U_{t+1}^1, \dots, U_{t+1}^\lambda \sim \mathcal{N}(0, I_d)$
- sort $f(m_t + \sigma_t C_t^{1/2} U_{t+1}^i)$

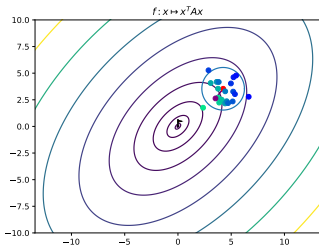


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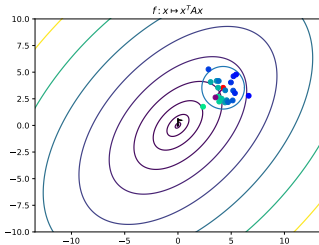


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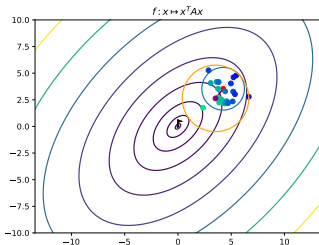


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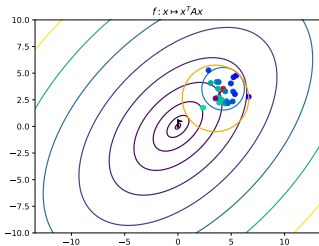


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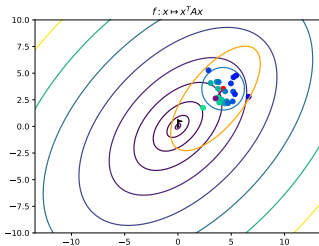


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Algorithm 2 d-CMA-ES

Given: $m_0 \in \mathbb{R}^d$, $\sigma_0 > 0$, $C_0 \in \mathcal{S}_{++}^d$, $p_0^\sigma \in \mathbb{R}^d$, $p_0^c \in \mathbb{R}^d$

For $t = 0, 1, 2, \dots$:

- $U_{t+1}^1, \dots, U_{t+1}^\lambda \sim \mathcal{N}(0, I_d)$
 - sort $f(m_t + \sigma_t C_t^{1/2} U_{t+1}^i)$
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Given: $m_0 \in \mathbb{R}^d$, $\sigma_0 > 0$, $C_0 \in \mathcal{S}_{++}^d$, $p_0^\sigma \in \mathbb{R}^d$, $p_0^c \in \mathbb{R}^d$

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 - $m_{t+1} = m_t + \sigma_t C_t^{1/2} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
 - $p_{t+1}^\sigma = (1 - c_\sigma) p_t^\sigma + \alpha_{c_\sigma} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
 - $\sigma_{t+1} = \sigma_t \exp\left(\frac{c_\sigma}{2d_\sigma} \left(\frac{\|p_{t+1}^\sigma\|^2}{d} - 1\right)\right)$ **smooth update**
 - $p_{t+1}^c = (1 - c_c) p_t^c + \alpha_{c_c} C_t^{1/2} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
 - $C_{t+1} = (1 - c_1 - c_\mu) C_t + c_1 (p_{t+1}^c)(p_{t+1}^c)^T + c_\mu \sum_{i=1}^\mu w_i \left(C_t^{1/2} U_{t+1}^{s_{t+1}(i)}\right) \left(C_t^{1/2} U_{t+1}^{s_{t+1}(i)}\right)^T$
-

One path variant of CMA-ES

Algorithm 3 a-CMA-ES

Given: $m_0 \in \mathbb{R}^d$, $\sigma_0 > 0$, $C_0 \in \mathcal{S}_{++}^d$, $p_0^\sigma \in \mathbb{R}^d$, $p_0^c \in \mathbb{R}^d$

For $t = 0, 1, 2, \dots$:

- $U_{t+1}^1, \dots, U_{t+1}^\lambda \sim \mathcal{N}(0, I_d)$
 - sort $f(m_t + \sigma_t C_t^{1/2} U_{t+1}^i)$
 - $m_{t+1} = m_t + \sigma_t C_t^{1/2} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
 - ~~$p_{t+1}^\sigma = (1 - c_\sigma) p_t^\sigma + \alpha_{c_\sigma} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$~~
 - ~~$\sigma_{t+1} = \sigma_t \exp\left(\frac{c_\sigma}{d_\sigma} \left(\frac{\|p_{t+1}^\sigma\|}{\mathbb{E}\|\mathcal{N}(0, I_d)\|} - 1\right)\right)$~~
 - $p_{t+1}^c = (1 - c_c) p_t^c + \alpha_{c_c} C_t^{1/2} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
 - $C_{t+1} = (1 - c_1 - c_\mu) C_t + c_1 (p_{t+1}^c) (p_{t+1}^c)^T + c_\mu \sum_{i=1}^\mu w_i \left(C_t^{1/2} U_{t+1}^{s_{t+1}(i)}\right) \left(C_t^{1/2} U_{t+1}^{s_{t+1}(i)}\right)^T$
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One path variant of CMA-ES

Algorithm 3 a-CMA-ES

Given: $m_0 \in \mathbb{R}^d$, $\sigma_0 > 0$, $C_0 \in \mathcal{S}_{++}^d$, $p_0^\sigma \in \mathbb{R}^d$, $p_0^c \in \mathbb{R}^d$

For $t = 0, 1, 2, \dots$:

- $U_{t+1}^1, \dots, U_{t+1}^\lambda \sim \mathcal{N}(0, I_d)$
 - sort $f(m_t + \sigma_t C_t^{1/2} U_{t+1}^i)$
 - $m_{t+1} = m_t + \sigma_t C_t^{1/2} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
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 - $p_{t+1}^c = (1 - c_c) p_t^c + \alpha_{c_c} C_t^{1/2} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
 - $C_{t+1} = (1 - c_1 - c_\mu) C_t + c_1 (p_{t+1}^c)(p_{t+1}^c)^T + c_\mu \sum_{i=1}^\mu w_i \left(C_t^{1/2} U_{t+1}^{s_{t+1}(i)}\right) \left(C_t^{1/2} U_{t+1}^{s_{t+1}(i)}\right)^T$
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One path variant of CMA-ES

Algorithm 3 a-CMA-ES

Given: $m_0 \in \mathbb{R}^d$, $\sigma_0 > 0$, $C_0 \in \mathcal{S}_{++}^d$, $p_0^\sigma \in \mathbb{R}^d$, $p_0^c \in \mathbb{R}^d$

For $t = 0, 1, 2, \dots$:

- $U_{t+1}^1, \dots, U_{t+1}^\lambda \sim \mathcal{N}(0, I_d)$
 - sort $f(m_t + \sigma_t C_t^{1/2} U_{t+1}^i)$
 - $m_{t+1} = m_t + \sigma_t C_t^{1/2} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
 - ~~$p_{t+1}^\sigma = (1 - c_\sigma) p_t^\sigma + \alpha_{c_\sigma} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$~~
 - ~~$\sigma_{t+1} =$~~
 $\sigma_t \exp\left(\frac{c_\sigma}{d_\sigma} \left(\frac{\|C_t^{-1/2} p_{t+1}^c\|}{\mathbb{E}\|\mathcal{N}(0, I_d)\|} - 1\right)\right)$ **only one path is used**
 - $p_{t+1}^c = (1 - c_c) p_t^c + \alpha_{c_c} C_t^{1/2} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
 - $C_{t+1} = (1 - c_1 - c_\mu) C_t + c_1 (p_{t+1}^c)(p_{t+1}^c)^T$
 $+ c_\mu \sum_{i=1}^\mu w_i \left(C_t^{1/2} U_{t+1}^{s_{t+1}(i)}\right) \left(C_t^{1/2} U_{t+1}^{s_{t+1}(i)}\right)^T$
-

CMA-ES without cumulation on the stepsize

Algorithm 4 s-CMA-ES

Given: $m_0 \in \mathbb{R}^d$, $\sigma_0 > 0$, $C_0 \in \mathcal{S}_{++}^d$, $p_0^\sigma \in \mathbb{R}^d$, $p_0^c \in \mathbb{R}^d$

For $t = 0, 1, 2, \dots$:

- $U_{t+1}^1, \dots, U_{t+1}^\lambda \sim \mathcal{N}(0, I_d)$
 - sort $f(m_t + \sigma_t C_t^{1/2} U_{t+1}^i)$
 - $m_{t+1} = m_t + \sigma_t C_t^{1/2} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
 - $p_{t+1}^\sigma = \cancel{(1 - c_\sigma) p_t^\sigma} + \alpha_{c_\sigma} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
 - $\sigma_{t+1} = \sigma_t \exp\left(\frac{c_\sigma}{d_\sigma} \left(\frac{\|p_{t+1}^\sigma\|}{\mathbb{E}\|\mathcal{N}(0, I_d)\|} - 1\right)\right)$
 - $p_{t+1}^c = (1 - c_c) p_t^c + \alpha_{c_c} C_t^{1/2} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
 - $C_{t+1} = (1 - c_1 - c_\mu) C_t + c_1 (p_{t+1}^c)(p_{t+1}^c)^T + c_\mu \sum_{i=1}^\mu w_i \left(C_t^{1/2} U_{t+1}^{s_{t+1}(i)}\right) \left(C_t^{1/2} U_{t+1}^{s_{t+1}(i)}\right)^T$
-

Algorithm 4 s-CMA-ES

Given: $m_0 \in \mathbb{R}^d$, $\sigma_0 > 0$, $C_0 \in \mathcal{S}_{++}^d$, $p_0^\sigma \in \mathbb{R}^d$, $p_0^c \in \mathbb{R}^d$

For $t = 0, 1, 2, \dots$:

- $U_{t+1}^1, \dots, U_{t+1}^\lambda \sim \mathcal{N}(0, I_d)$
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Given: $m_0 \in \mathbb{R}^d$, $\sigma_0 > 0$, $C_0 \in \mathcal{S}_{++}^d$, $p_0^\sigma \in \mathbb{R}^d$, $p_0^c \in \mathbb{R}^d$

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 - $m_{t+1} = m_t + \sigma_t C_t^{1/2} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
 - $p_{t+1}^\sigma = \alpha_1 \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$ **no cumulation on the stepsize**
 - $\sigma_{t+1} = \sigma_t \exp\left(\frac{c_\sigma}{d_\sigma} \left(\frac{\|p_{t+1}^\sigma\|}{\mathbb{E}\|\mathcal{N}(0, I_d)\|} - 1\right)\right)$
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-

Algorithm 5 d-CMA-ES

Given: $m_0 \in \mathbb{R}^d$, $\sigma_0 > 0$, $C_0 \in \mathcal{S}_{++}^d$, $p_0^\sigma \in \mathbb{R}^d$, $p_0^c \in \mathbb{R}^d$

For $t = 0, 1, 2, \dots$:

- $U_{t+1}^1, \dots, U_{t+1}^\lambda \sim \mathcal{N}(0, I_d)$
 - sort $f(m_t + \sigma_t C_t^{1/2} U_{t+1}^i)$
 - $m_{t+1} = m_t + \sigma_t C_t^{1/2} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
 - $p_{t+1}^\sigma = (1 - c_\sigma) p_t^\sigma + \alpha_{c_\sigma} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
 - $\sigma_{t+1} = \sigma_t \exp\left(\frac{c_\sigma}{d_\sigma} \left(\frac{\|p_{t+1}^\sigma\|}{\mathbb{E}\|\mathcal{N}(0, I_d)\|} - 1\right)\right)$
 - ~~$p_{t+1}^c = (1 - c_c) p_t^c$~~ $p_{t+1}^c = \alpha_{c_c} C_t^{1/2} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$
 - $C_{t+1} = (1 - c_1 - c_\mu) C_t + c_1 (p_{t+1}^c)(p_{t+1}^c)^T + c_\mu \sum_{i=1}^\mu w_i \left(C_t^{1/2} U_{t+1}^{s_{t+1}(i)}\right) \left(C_t^{1/2} U_{t+1}^{s_{t+1}(i)}\right)^T$
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Given: $m_0 \in \mathbb{R}^d$, $\sigma_0 > 0$, $C_0 \in \mathcal{S}_{++}^d$, $p_0^\sigma \in \mathbb{R}^d$, $p_0^c \in \mathbb{R}^d$

For $t = 0, 1, 2, \dots$:

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Given: $m_0 \in \mathbb{R}^d$, $\sigma_0 > 0$, $C_0 \in \mathcal{S}_{++}^d$, $p_0^\sigma \in \mathbb{R}^d$, $p_0^c \in \mathbb{R}^d$

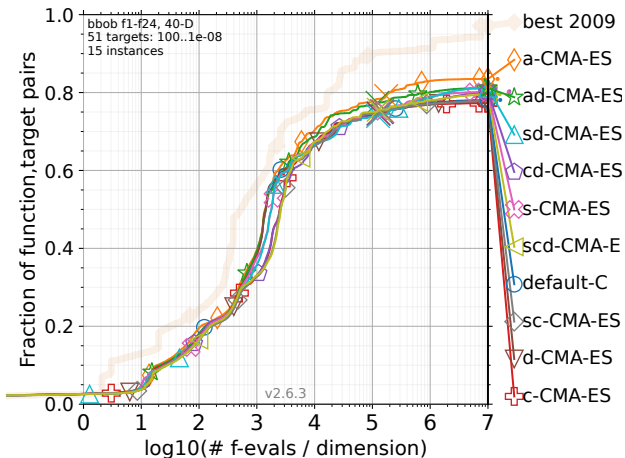
For $t = 0, 1, 2, \dots$:

- $U_{t+1}^1, \dots, U_{t+1}^\lambda \sim \mathcal{N}(0, I_d)$
 - sort $f(m_t + \sigma_t C_t^{1/2} U_{t+1}^i)$
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 - $p_{t+1}^c = \alpha_1 C_t^{1/2} \sum_{i=1}^\mu w_i U_{t+1}^{i:\lambda}$ **no cumulation on C_t**
 - $C_{t+1} = (1 - c_1 - c_\mu) C_t + c_1 (p_{t+1}^c)(p_{t+1}^c)^T + c_\mu \sum_{i=1}^\mu w_i \left(C_t^{1/2} U_{t+1}^{s_{t+1}(i)}\right) \left(C_t^{1/2} U_{t+1}^{s_{t+1}(i)}\right)^T$
-

10 variants of CMA-ES

- default-CMA-ES
- d-CMA-ES: with a smooth update on the stepsize
- a-CMA-ES: using one path for both cumulations
- s-CMA-ES: no cumulation on the stepsize
- c-CMA-ES: no cumulation on the covariance matrix
- ad-CMA-ES, sd-CMA-ES, cd-CMA-ES, sc-CMA-ES, scd-CMA-ES

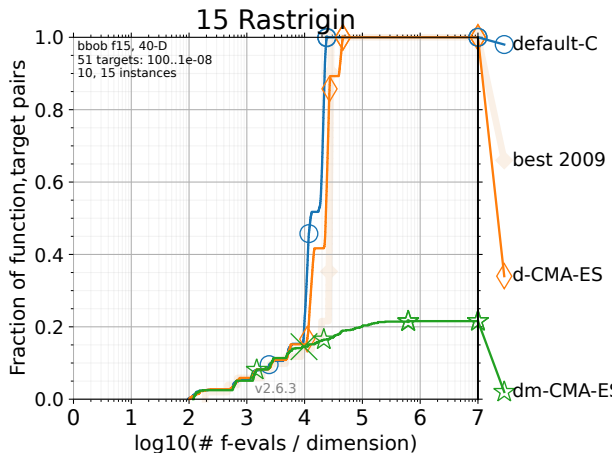
10 variants of CMA-ES



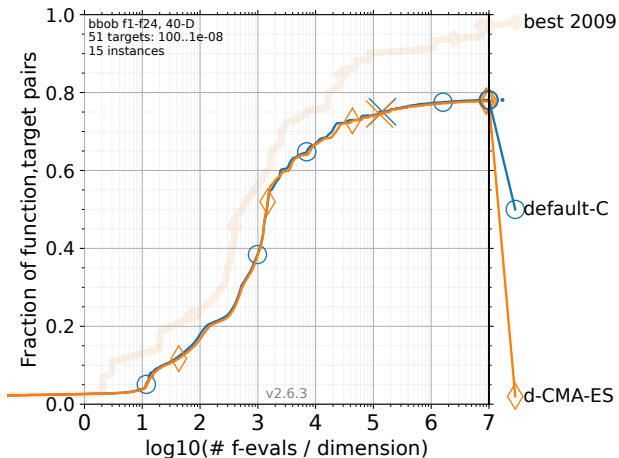
10 variants of CMA-ES

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d-CMA-ES: the parameter d_σ has to be scaled correctly



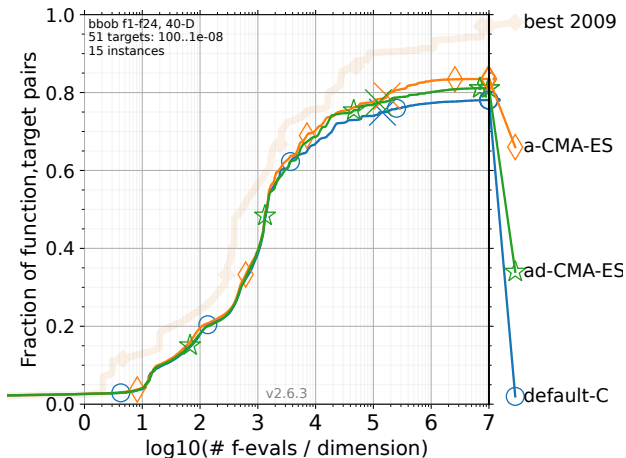
d-CMA-ES: no significant differences



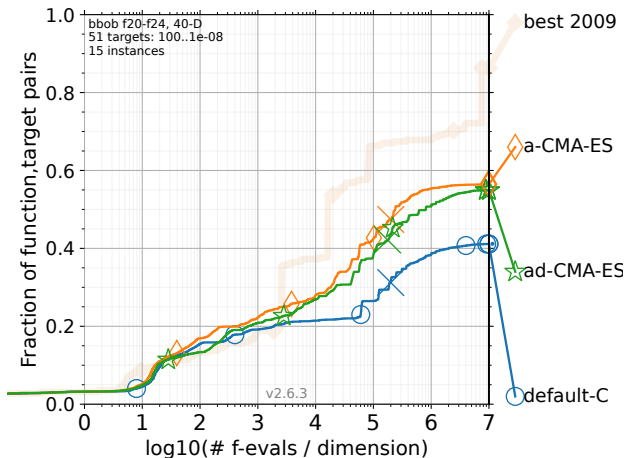
10 variants of CMA-ES

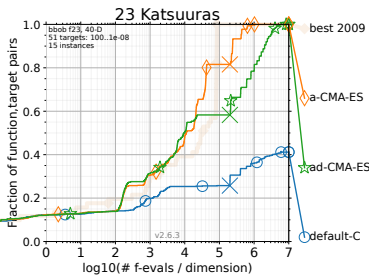
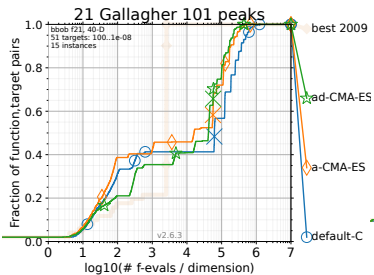
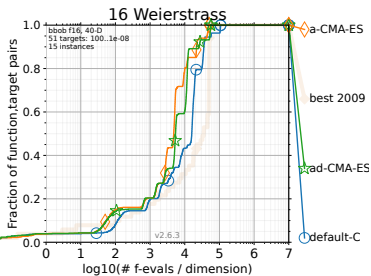
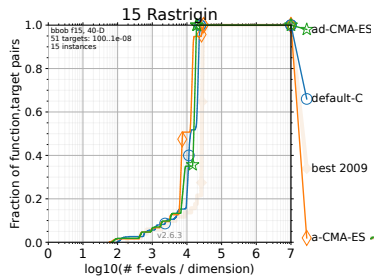
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a-CMA-ES



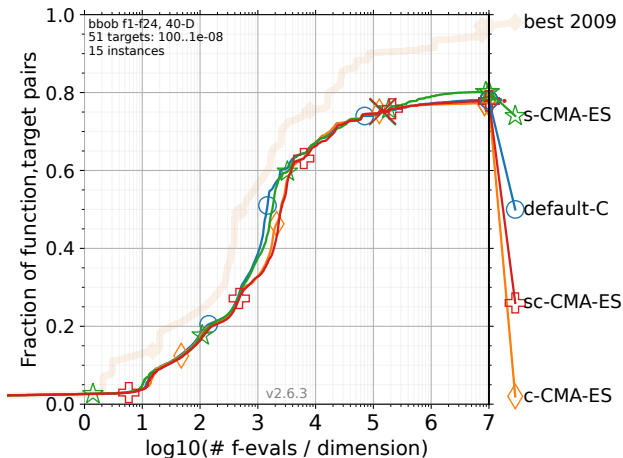
a-CMA-ES

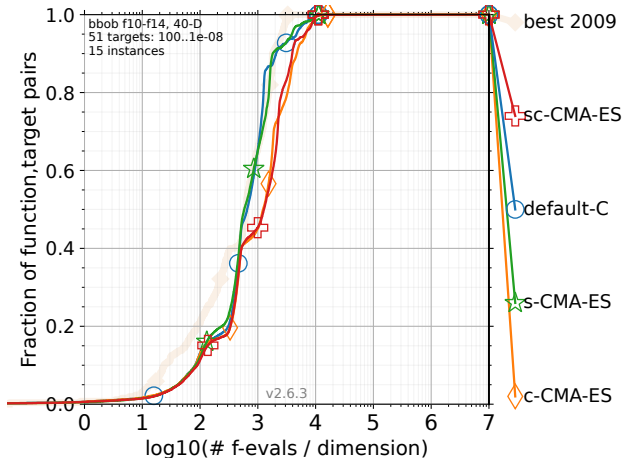


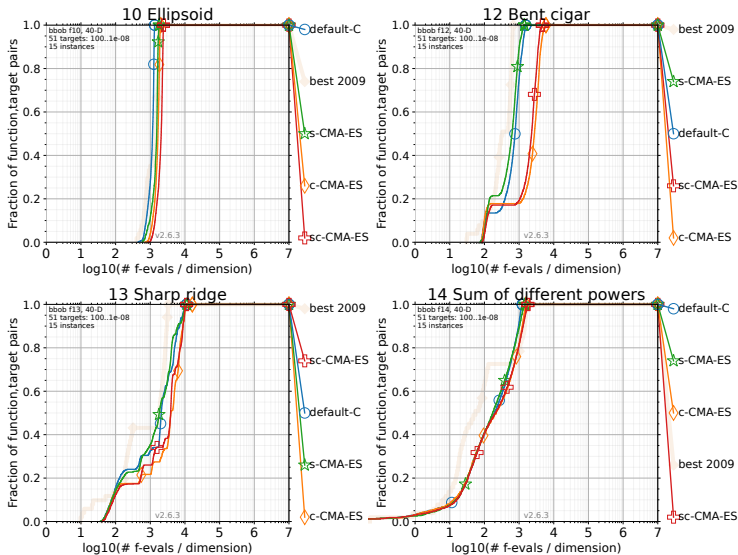


10 variants of CMA-ES

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scd-CMA-ES







Conclusions

- For **theoretical** purposes, we introduce several modifications to CMA-ES
- None of them **break** the algorithm – in the sense that the algorithm still converges to the solution of the problems it already solved
- **Cumulation on the covariance matrix** is important for the performances of the algorithm on highly **ill-conditioned** problems
- Otherwise there is no significantly remarkable loss of performances due to these modifications

Thank you!