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Armand Gissler, Anne Auger, Nikolaus Hansen

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Supplementary Material for Learning Rate Adaptation by Line Search in Evolution Strategies with Recombination

Armand Gissler forename.lastname@polytechnique.edu Inria and CMAP, Ecole Polytechnique, Inria and CMAP, Ecole Polytechnique, IP Paris, CNRS Palaiseau, France

Anne Auger forename.lastname@inria.fr IP Paris, CNRS Palaiseau, France

Nikolaus Hansen forename.lastname@inria.fr Inria and CMAP, Ecole Polytechnique, IP Paris, CNRS Palaiseau, France

ABSTRACT

This is the supplementary material (proofs) for [1].

PROOF OF LEMMA 2.3 IN [1]. Let $t \in \mathbb{N}$. We know that X_{t+1} = $X_t + \kappa_{t+1} \sum_{i=1}^{\mu} w_i U_{t+1}^{\varphi(i)}$ $\mathcal{L}_{t+1}^{\varphi(i)}$, with $\kappa_{t+1} = \bar{\kappa}(X_t, \alpha || X_t || \sum_{i=1}^{\mu} w_i U_{t+1}^{\varphi(i)}$ $_{t+1}^{(\varphi(i))}$. Then, by (A1)

$$
Z_t = \frac{||X_{t+1}||}{||X_t||} = \left\| \frac{X_t}{||X_t||} + \alpha \bar{\kappa} \left(\frac{X_t}{||X_t||}, \alpha S_{U_{t+1}}^{\varphi} \right) S_{U_{t+1}}^{\varphi} \right\|.
$$

But then, according to (A2), we have that Z_t has the same distribution than $||e_1 + \alpha \bar{\kappa}(e_1, \alpha S_V^{\varphi}) S_V^{\varphi}$ $\mathcal{L}_{V}^{\varphi}$ ||. Thus, all the $Z_t, t \in \mathbb{N}$ are identically distributed.

Let $s \in \mathbb{R}$, and denote $\psi^s : \mathbb{R}^n \to \mathbb{C}$ the measurable function such that $\psi^s(X_t) = \mathbb{E}[\exp isZ_t | X_t].$ Then, for $x \in \mathbb{R}$, according to (A2) applied to $X = x$

$$
\psi^s(x) = \mathbb{E} \exp is \left\| \frac{x}{\|x\|} + \alpha \bar{\kappa} \left(\frac{x}{\|x\|}, \alpha S_{U_{t+1}}^{\varphi} \right) S_{U_{t+1}}^{\varphi} \right\|
$$

= $\psi^s(e_1).$

Thus, $\mathbb{E}[\exp isZ_t | X_t] = \psi^s(X_t) = \psi^s(e_1)$. Define \mathcal{F}_t the filtration induced by X_0, \ldots, X_t . We get then

$$
\mathbb{E}[\exp i(s_0Z_0 + s_1Z_1 + \dots + s_tZ_t)]
$$
\n
$$
= \mathbb{E}[\mathbb{E}[\exp i(s_0Z_0 + s_1Z_1 + \dots + s_tZ_t) | \mathcal{F}_t]]
$$
\n
$$
= \mathbb{E}[\exp is(s_0Z_0 + s_1Z_1 + \dots + s_{t-1}Z_{t-1}) \times \mathbb{E}[\exp is_tZ_t | X_t]]
$$
\n
$$
= \mathbb{E}[\exp i(s_0Z_0 + s_1Z_1 + \dots + s_{t-1}Z_{t-1})] \times \psi^{s_t}(e_1)
$$
\n
$$
= \mathbb{E}[\exp i(s_0Z_0 + s_1Z_1 + \dots + s_{t-1}Z_{t-1})] \times \mathbb{E}[\exp is_tZ_t]
$$

Hence, Z_t is independent from $(Z_0, Z_1, \ldots, Z_{t-1})$. By induction, we get that (Z_0, Z_1, \ldots, Z_t) is mutually independent. □

PROOF OF LEMMA 3.2 IN [1]. Assumption (A1) holds by [1, Lemma 2.1].

Let *R* be a rotation matrix. Then $\bar{k}_{\text{PLS}}(Rx, Rv) = \arg \min_{\kappa \geq 0} ||Rx +$ $\kappa Rv \|\ = \arg \min_{\kappa \geq 0} \|R(x + \kappa v)\| = \arg \min_{\kappa \geq 0} \|x + \kappa v\| = \overline{\kappa_{\text{PLS}}}(x, v).$ Hence $\bar{\kappa}_{PI,S}$ is rotation-invariant and [1, Lemma 2.2] implies that (A2) holds.

Now, let us show that (A4) is satisfied. Let $U^1, \ldots, U^{\lambda} \in \mathbb{R}^{\mathbb{N}}$ be i.i.d. infinite dimensional standard Gaussian vectors. Then, when $n \rightarrow \infty$, the following limit holds a.s. by the strong LLN

$$
\frac{1}{n} \left\| \sum_{i=1}^{\mu} w_i [U^i]_{\le n} \right\|^2 = \frac{1}{n} \sum_{k=1}^n \left| \sum_{i=1}^{\mu} w_i [U^i]_k \right|^2 \to \mu_w^{-1},
$$

thus by [1, Lemma 3.1]

$$
\bar{\kappa}_{\rm PLS} \left(e_1, \frac{\alpha}{n} \sum_{i=1}^{\mu} w_i [U^i]_{\leq n} \right) = -\mathbf{1}_{\sum_{i=1}^{\mu} w_i [U^i]_1 < 0} \frac{\sum_{i=1}^{\mu} w_i [U^i]_1}{\frac{\alpha}{n} \|\sum_{i=1}^{\mu} w_i [U^i]_{\leq n} \|^2} \to -\mathbf{1}_{\sum_{i=1}^{\mu} w_i [U^i]_1 < 0} \alpha^{-1} \mu_{\rm w} \sum_{i=1}^{\mu} w_i [U^i]_1.
$$

Thus $(A4)$ holds. \square

Proof of Lemma 4.2 in [1]. First, let us prove the scaling-invariance condition (A1) for $\bar{\kappa}_{\text{DLS}}^{\varepsilon,\beta}$. Consider $\bar{\kappa}_{\text{DLS}}^{\varepsilon,\beta}(x/r,v/r,\kappa_{\text{init}})$. Then, as in line 4 of [1, Algorithm 2], the condition $f(x + \kappa^0 v) < f(x + \kappa^1 v)$ is equivalent to $f(x/r + \kappa^0 v/r) < f(x/r + \kappa^1 v/r)$, as f is the sphere function is scaling-invariant, then $\bar{\kappa}_{\text{DLS}}^{ \varepsilon,\beta}(x/r,v/r,\kappa_{\text{init}})$ = $\bar{\kappa}_{\rm DLS}^{\varepsilon,\beta}(x,v,\kappa_{\rm init})$. Thus, (A1) holds.

The function $\bar{\kappa}_{\text{DLS}}^{\varepsilon,\beta}$ is rotation-invariant. Indeed if R is a rotation matrix, then, as in line 4 of [1, Algorithm 2], the condition $f(x +$ $\kappa^{0}v$) < $f(x+\kappa^{1}v)$ is equivalent to $\widehat{f}(Rx+\kappa^{0}Rv) < f(Rx+\kappa^{1}Rv)$, as \boldsymbol{f} is the sphere function is invariant by rotation, then this implies that $\bar{\kappa}_{\text{DLS}}^{\varepsilon,\beta}(Rx,Rv,\kappa_{\text{init}}) = \bar{\kappa}_{\text{DLS}}^{\varepsilon,\beta}(x,v,\kappa_{\text{init}})$. Thus, by [1, Lemma 2.2], (A2) holds.

We prove now that (A4) is satisfied. Consider $U^1, \ldots, U^{\mu} \in \mathbb{R}^{\mathbb{N}}$ μ i.i.d. infinite dimensional standard Gaussian vectors, and denote $S_U^n = \sum_{i=1}^{\mu} w_i [U^i]_{\le n}.$

Consider the line search obtained with $\bar{\kappa}_{\rm DLS}^{\varepsilon,\beta}\left(e_1, \frac{\alpha}{n} S_U^n, \kappa_{\rm init}\right)$. We denote $(\kappa^{i,0}, \kappa^{i,1})_{i=0,\cdots,C(\beta,\varepsilon)-2}$ the value of κ^1 and κ^0 over the iterations of this line search.

We prove now by induction that for all $i = 0, 1, \ldots$, the following limits when $n \to \infty$ hold a.s. $\kappa^{i,0} \to \kappa^{\infty,i,0}$ and $\kappa^{i,1} \to \kappa^{\infty,i,1}$, where $\kappa^{\infty,i,0}, \kappa^{\infty,i,1}$ are given in the line 5 and line 7 of [1, Algorithm 2] over iteration initialized with parameters $X = \alpha^{-1} \mu_w S_U^{\bar{1}} \in \mathbb{R}^1$, $v =$ 1 ∈ \mathbb{R}^1 , and given κ_{init} , ε , β . At iteration $i = 0$, this is trivially true as $\kappa^{i,0} = \kappa^{\infty,i,0} = \kappa_{\text{init}}/2$, and $\kappa^{i,1} = \kappa^{\infty,i,1} = 2\kappa_{\text{init}}$

Now consider an arbitrary step *i* of the line search, and assume that $\kappa^{i,0} \to \kappa^{\infty,i,0}$ and $\kappa^{i,1} \to \kappa^{\infty,i,1}$. Then, $\kappa^{i+1,0} = \kappa^{i,0} + (1 \beta$) $(\kappa^{i,1} - \kappa^{i,0})$ 1 $_{\{h_{\alpha/n}(\kappa^{i,1}S_U^n) < h_{\alpha/n}(\kappa^{i,0}S_U^n)\}}$, where

$$
h_{\alpha/n}(\kappa^{i,j} S_U^n) = 2 \underbrace{\kappa^{i,j}}_{\to \kappa^{\infty, i,j}} S_U^1 + \alpha \left(\kappa^{i,j}\right)^2 \underbrace{\frac{\|S_U^n\|^2}{n}}_{\to \sum_{i=1}^n w_i^2}
$$

Hence, a.s. $h_{\alpha/n}(\kappa^{i,j}S_U^n)$ tends to $2\kappa^{\infty,i,j}S_U^1 + \alpha (\kappa^{\infty,i,j})^2 \mu_{\rm w}^{-1}$ when $n \rightarrow \infty$, so that

$$
\lim_{n \to \infty} \mathbf{1}_{\{h_{\alpha/n}(\kappa^{i,1} S_U^n) < h_{\alpha/n}(\kappa^{i,0} S_U^n)\}} \\
= \mathbf{1}_{\{(\kappa^{\infty,i,1} - \kappa^{\infty,i,0}) \Big(2S_U^n + (\kappa^{\infty,i,0} + \kappa^{\infty,i,1}) \alpha \sum_{i=1}^\mu w_i^2\Big) > 0\}},
$$

hence $\kappa^{i+1,0} \to \kappa^{\infty, i+1,0}$. Similarly, $\lim_{n \to \infty} \kappa^{i+1,1} = \kappa^{\infty, i+1,1}$.

In the end, by induction, we get that

$$
\lim_{n \to \infty} \bar{\kappa} \left(e_1, \frac{\alpha}{n} S_U^n, \kappa_{\text{init}} \right) = \bar{\kappa}^{\infty} \left(\left(\begin{bmatrix} U^i \end{bmatrix}_1 \right)_{i=1,\dots,\lambda}, \kappa_{\text{init}} \right)
$$

where $\bar{\kappa}^{\infty}$ is defined in [1, Lemma 4.2]. Hence (A4) holds.

 \Box

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LEMMA A.1. For $n \in \mathbb{N}^*$, let U_n^1, \ldots, U_n^μ be i.i.d. standard multivariate normal distribution of dimension n. Suppose that the functions $\bar{\kappa}_n : \mathbb{R}^n \times \mathbb{R}^n \to \mathbb{R}_+$ are either all equal to $\bar{\kappa}_{\text{PLS}}$ or all upperbounded by a positive constant κ^M . Then,

$$
\left(n\ln\left\|e_1+\bar{\kappa}_n\left(e_1,\alpha/n\sum_{i=1}^\mu w_iU_n^i\right)\alpha/n\sum_{i=1}^n w_iU_n^i\right\|^2\right)_{n\in\mathbb{N}^*}
$$

is uniformly integrable.

Proof. Suppose first that for all $n \in \mathbb{N}^*$, $\bar{\kappa}_n = \bar{\kappa}_{\text{PLS}}$. Note then that, as in the proof of [1, Theorem 3.3], we have

$$
n \ln \left\| e_1 + \bar{\kappa}_n \left(e_1, \alpha / n \sum_{i=1}^{\mu} w_i U_n^i \right) \frac{\alpha}{n} \sum_{i=1}^n w_i U_n^i \right\|^2
$$

=
$$
-8 \ln \left\{ \left(1 - \frac{\left[\sum_{i=1}^{\mu} w_i U_n^i \right]_1^2}{\left\| \sum_{i=1}^{\mu} w_i U_n^i \right\|^2} \right)^{-n/8} \right\} \mathbf{1}_{\left[\sum_{i=1}^{\mu} w_i U_n^i \right]_1 < 0}.
$$

But it is proven in [2, Proof of Proposition 4, page 23] that

$$
\left(\ln\left[\left(1-\frac{\left[\sum_{i=1}^{\mu}w_iU_n^i\right]_1^2}{\left\|\sum_{i=1}^{\mu}w_iU_n^i\right\|^2}\right)^{-n/8}\right]1_{\left[\sum_{i=1}^{\mu}w_iU_n^i\right]_1<0}\right)_{n\in\mathbb{N}^*}
$$

is uniformly integrable.

Suppose now that there exists $\kappa^M > 0$ such that for all $n \in \mathbb{N}^*$, $\bar{\kappa}_n \leq \kappa^M$. We will prove the uniform integrability of the positive part and of the negative part of

$$
\left(n\ln\left\|e_1+\bar{\kappa}_n\left(e_1,\alpha/n\sum_{i=1}^\mu w_iU_n^i\right)\alpha/n\sum_{i=1}^n w_iU_n^i\right\|^2\right)_{n\in\mathbb{N}^*}
$$

For the negative part, note that, by definition of $\bar{\kappa}_{\rm PLS}$, then

$$
\begin{aligned} \left\|e_1 + \bar{\kappa}_n\left(e_1, \alpha/n\sum_{i=1}^\mu w_i U_n^i\right)\alpha/n\sum_{i=1}^n w_i U_n^i\right\|^2\\ \geqslant \left\|e_1 + \bar{\kappa}_{\text{PLS}}\left(e_1, \alpha/n\sum_{i=1}^\mu w_i U_n^i\right)\alpha/n\sum_{i=1}^n w_i U_n^i\right\|^2. \end{aligned}
$$

In addition, the negative part of the logarithm \ln^{-1} is a decreasing function, thus

$$
n \ln^{-} \left\| e_1 + \bar{\kappa}_n \left(e_1, \alpha/n \sum_{i=1}^{\mu} w_i U_n^i \right) \alpha/n \sum_{i=1}^n w_i U_n^i \right\|^2
$$

\$\leq n \ln^{-} \left\| e_1 + \bar{\kappa}_{\text{PLS}} \left(e_1, \alpha/n \sum_{i=1}^{\mu} w_i U_n^i \right) \alpha/n \sum_{i=1}^n w_i U_n^i \right\|^2 \qquad (1)

Note that, as $||e_1 + \bar{\kappa}_{\text{PLS}}(e_1, \alpha/n \sum_{i=1}^{\mu} w_i U_n^i) \alpha/n \sum_{i=1}^n w_i U_n^i|| \le ||e_1 +$ 0 $\|\hat{v}_1\| = 1$, then $\ln \|e_1 + \bar{k}_{PLS}(e_1, \alpha/n \sum_{i=1}^{\mu} w_i U_n^i) \alpha/n \sum_{i=1}^{\mu} w_i U_n^i\| \leq$ $||e_1 + 0|| \le 0$, thus is equal in absolute value to its negative part. However, as

$$
\left(n \ln^{-1} \left\| e_1 + \bar{\kappa}_{\text{PLS}} \left(e_1, \alpha/n \sum_{i=1}^{\mu} w_i U_n^i \right) \alpha/n \sum_{i=1}^n w_i U_n^i \right\|^2 \right)_{n \in \mathbb{N}^*}
$$

is uniformly integrable (since it is equal to the sequence with \ln

instead of ln[−] which is uniformly integrable as seen above) , then according to Eq. (1) so is

$$
\left(n\ln^{-1}\left\|e_1+\bar{\kappa}_n\left(e_1,\alpha/n\sum_{i=1}^\mu w_iU_n^i\right)\alpha/n\sum_{i=1}^n w_iU_n^i\right\|^2\right)_{n\in\mathbb{N}^*}.
$$

We claim that for all $x \in \mathbb{R}^n$, and $0 \le \kappa \le \kappa^M$, then

$$
\ln^+ \|e_1+\kappa x\|\leqslant \ln^+ \|e_1+\kappa^M x\|.
$$

Indeed, the above equation is equivalent to

$$
||e_1 + \kappa x||^2 \mathbf{1}_{||e_1 + \kappa x|| \ge 1} \le ||e_1 + \kappa^M x||^2 \mathbf{1}_{||e_1 + \kappa^M x|| \ge 1}.
$$

However, the derivative of the function $\kappa \mapsto \|e_1 + \kappa x\|^2$ is equal to $\kappa \mapsto 2[x]_1 + \kappa \langle x, x \rangle$, hence is nonnegative for any $\kappa \geq 0$ that also satisfies that $||e_1 + \kappa x||^2 = 1 + \kappa(2[x]_1 + \kappa \langle x, x \rangle)$ is greater than or equal to 1. Thus, the above condition is satisfied.

For the positive part, we have then

$$
n \ln^{+} \left\| e_1 + \bar{\kappa}_n \left(e_1, \frac{\alpha}{n} \sum_{i=1}^{\mu} w_i U_n^i \right) \frac{\alpha}{n} \sum_{i=1}^{\mu} w_i U_n^i \right\|^2
$$

$$
\leq n \ln^{+} \left\| e_1 + \frac{\kappa^M \alpha}{n} \sum_{i=1}^{\mu} w_i U_n^i \right\|^2.
$$

But it is proven in $[2]^1$ that the RHS of the above equation is uniformly integrable. All in all, we get that

$$
\left(n\ln\left\|e_1+\bar{\kappa}_n\left(e_1,\alpha/n\sum_{i=1}^{\mu}w_iU_n^i\right)\alpha/n\sum_{i=1}^n w_iU_n^i\right\|^2\right)_{n\in\mathbb{N}^*}
$$

is uniformly integrable.

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¹ replace σ^* by $\alpha \kappa^M$ in the proof of Proposition 4 of [2]