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

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#### Abstract

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


Proof of Lemma 2.3 in [1]. Let $t \in \mathbb{N}$. We know that $X_{t+1}=$ $X_{t}+\kappa_{t+1} \sum_{i=1}^{\mu} w_{i} U_{t+1}^{\varphi(i)}$, with $\kappa_{t+1}=\bar{\kappa}\left(X_{t}, \alpha\left\|X_{t}\right\| \sum_{i=1}^{\mu} w_{i} U_{t+1}^{\varphi(i)}\right)$. Then, by (A1)

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

But then, according to (A2), we have that $Z_{t}$ has the same distribution than $\left\|e_{1}+\alpha \bar{\kappa}\left(e_{1}, \alpha S_{V}^{\varphi}\right) S_{V}^{\varphi}\right\|$. Thus, all the $Z_{t}, t \in \mathbb{N}$ are identically distributed.

Let $s \in \mathbb{R}$, and denote $\psi^{s}: \mathbb{R}^{n} \rightarrow \mathbb{C}$ the measurable function such that $\psi^{s}\left(X_{t}\right)=\mathbb{E}\left[\exp i s Z_{t} \mid X_{t}\right]$. Then, for $x \in \mathbb{R}$, according to (A2) applied to $X=x$

$$
\begin{aligned}
\psi^{s}(x) & =\mathbb{E} \exp i s\left\|\frac{x}{\|x\|}+\alpha \bar{\kappa}\left(\frac{x}{\|x\|}, \alpha S_{U_{t+1}}^{\varphi}\right) S_{U_{t+1}}^{\varphi}\right\| \\
& =\psi^{s}\left(e_{1}\right) .
\end{aligned}
$$

Thus, $\mathbb{E}\left[\exp\right.$ is $\left.Z_{t} \mid X_{t}\right]=\psi^{s}\left(X_{t}\right)=\psi^{s}\left(e_{1}\right)$. Define $\mathcal{F}_{t}$ the filtration induced by $X_{0}, \ldots, X_{t}$. We get then

$$
\begin{aligned}
\mathbb{E} & {\left[\exp i\left(s_{0} Z_{0}+s_{1} Z_{1}+\cdots+s_{t} Z_{t}\right)\right] } \\
& =\mathbb{E}\left[\mathbb{E}\left[\exp i\left(s_{0} Z_{0}+s_{1} Z_{1}+\cdots+s_{t} Z_{t}\right) \mid \mathcal{F}_{t}\right]\right] \\
& =\mathbb{E}\left[\exp i s\left(s_{0} Z_{0}+s_{1} Z_{1}+\cdots+s_{t-1} Z_{t-1}\right) \times \mathbb{E}\left[\exp i s_{t} Z_{t} \mid X_{t}\right]\right] \\
& =\mathbb{E}\left[\exp i\left(s_{0} Z_{0}+s_{1} Z_{1}+\cdots+s_{t-1} Z_{t-1}\right)\right] \times \psi^{s_{t}}\left(e_{1}\right) \\
& =\mathbb{E}\left[\exp i\left(s_{0} Z_{0}+s_{1} Z_{1}+\cdots+s_{t-1} Z_{t-1}\right)\right] \times \mathbb{E}\left[\exp i s_{t} Z_{t}\right]
\end{aligned}
$$

Hence, $Z_{t}$ is independent from $\left(Z_{0}, Z_{1}, \ldots, Z_{t-1}\right)$. By induction, we get that $\left(Z_{0}, Z_{1}, \ldots, Z_{t}\right)$ 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 $\bar{\kappa}_{\text {PLS }}(R x, R v)=\arg \min _{\kappa \geqslant 0} \| R x+$ $\kappa R v\left\|=\arg \min _{\kappa \geqslant 0}\right\| R(x+\kappa v)\left\|=\arg \min _{\kappa \geqslant 0}\right\| x+\kappa v \|=\bar{\kappa}_{\text {PLS }}(x, v)$. Hence $\bar{\kappa}_{\text {PLS }}$ 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

$$
\frac{1}{n}\left\|\sum_{i=1}^{\mu} w_{i}\left[U^{i}\right]_{\leqslant n}\right\|^{2}=\frac{1}{n} \sum_{k=1}^{n}\left|\sum_{i=1}^{\mu} w_{i}\left[U^{i}\right]_{k}\right|^{2} \rightarrow \mu_{\mathrm{w}}^{-1}
$$

thus by [1, Lemma 3.1]

$$
\begin{aligned}
& \bar{\kappa}_{\mathrm{PLS}}\left(e_{1}, \frac{\alpha}{n} \sum_{i=1}^{\mu} w_{i}\left[U^{i}\right]_{\leqslant n}\right)=-1 \sum_{i=1}^{\mu} w_{i}\left[U^{i}\right]_{1}<0 \frac{\sum_{i=1}^{\mu} w_{i}\left[U^{i}\right]_{1}}{\frac{\alpha}{n}\left\|\sum_{i=1}^{\mu} w_{i}\left[U^{i}\right]_{\leqslant n}\right\|^{2}} \\
& \rightarrow \mathbf{1}_{\sum_{i=1}^{\mu} w_{i}\left[U^{i}\right]_{1}<0} \alpha^{-1} \mu_{\mathrm{w}} \sum_{i=1}^{\mu} w_{i}\left[U^{i}\right]_{1} .
\end{aligned}
$$

Thus (A4) holds.

Proof of Lemma 4.2 in [1]. First, let us prove the scaling-invariance condition (A1) for $\bar{\kappa}_{\text {DLS }}^{\varepsilon, \beta}$. Consider $\bar{\kappa}_{\text {DLS }}^{\varepsilon, \beta}\left(x / r, v / r, \kappa_{\text {init }}\right)$. Then, as in line 4 of [1, Algorithm 2], the condition $f\left(x+\kappa^{0} v\right)<f\left(x+\kappa^{1} v\right)$ is equivalent to $f\left(x / r+\kappa^{0} v / r\right)<f\left(x / r+\kappa^{1} v / r\right)$, as $f$ is the sphere function is scaling-invariant, then $\bar{\kappa}_{\text {DLS }}^{\varepsilon, \beta}\left(x / r, v / r, \kappa_{\text {init }}\right)=$ $\bar{\kappa}_{\text {DLS }}^{\varepsilon, \beta}\left(x, v, \kappa_{\text {init }}\right)$. 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 $f(x+$ $\left.\kappa^{0} v\right)<f\left(x+\kappa^{1} v\right)$ is equivalent to $f\left(R x+\kappa^{0} R v\right)<f\left(R x+\kappa^{1} R v\right)$, as $f$ is the sphere function is invariant by rotation, then this implies that $\bar{\kappa}_{\text {DLS }}^{\varepsilon, \beta}\left(R x, R v, \kappa_{\text {init }}\right)=\bar{\kappa}_{\text {DLS }}^{\varepsilon, \beta}\left(x, v, \kappa_{\text {init }}\right)$. Thus, by [1, Lemma 2.2], (A2) holds.
We prove now that (A4) is satisfied. Consider $U^{1}, \ldots, U^{\mu} \in \mathbb{R}^{\mathbb{N}}$ $\mu$ i.i.d. infinite dimensional standard Gaussian vectors, and denote $S_{U}^{n}=\sum_{i=1}^{\mu} w_{i}\left[U^{i}\right] \leqslant n$.

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

We prove now by induction that for all $i=0,1, \ldots$, the following limits when $n \rightarrow \infty$ hold a.s. $\kappa^{i, 0} \rightarrow \kappa^{\infty, i, 0}$ and $\kappa^{i, 1} \rightarrow \kappa^{\infty, i, 1}$, 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 $X=\alpha^{-1} \mu_{\mathrm{w}} S_{U}^{1} \in \mathbb{R}^{1}, v=$ $1 \in \mathbb{R}^{1}$, and given $\kappa_{\text {init }}, \varepsilon, \beta$. 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 $\kappa^{i, 0} \rightarrow \kappa^{\infty, i, 0}$ and $\kappa^{i, 1} \rightarrow \kappa^{\infty, i, 1}$. Then, $\kappa^{i+1,0}=\kappa^{i, 0}+(1-$ $\beta)\left(\kappa^{i, 1}-\kappa^{i, 0}\right) 1_{\left\{h_{\alpha / n}\left(\kappa^{i, 1} S_{U}^{n}\right)<h_{\alpha / n}\left(\kappa^{i, 0} S_{U}^{n}\right)\right\}}$, where

$$
h_{\alpha / n}\left(\kappa^{i, j} S_{U}^{n}\right)=2 \underbrace{\kappa^{i, j}}_{\rightarrow \kappa^{\kappa,, i, j}} S_{U}^{1}+\alpha\left(\kappa^{i, j}\right)^{2} \underbrace{\frac{\left\|S_{U}^{n}\right\|^{2}}{n}}_{\rightarrow \sum_{i=1}^{\mu} w_{i}^{2}}
$$

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

$$
\begin{aligned}
\lim _{n \rightarrow \infty} 1_{\left\{h_{\alpha / n}\left(\kappa^{i, 1} S_{U}^{n}\right)<h_{\alpha / n}\left(\kappa^{i, 0} S_{U}^{n}\right)\right\}} \\
\quad=1_{\left\{\left(\kappa^{\infty, i, 1}-\kappa^{\infty, i, 0}\right)\left(2 S_{U}^{n}+\left(\kappa^{\infty, i, 0}+\kappa^{\infty, i, 1}\right) \alpha \sum_{i=1}^{\mu} w_{i}^{2}\right)>0\right\}}
\end{aligned}
$$

hence $\kappa^{i+1,0} \rightarrow \kappa^{\infty, i+1,0}$. Similarly, $\lim _{n \rightarrow \infty} \kappa^{i+1,1}=\kappa^{\infty, i+1,1}$.
In the end, by induction, we get that

$$
\lim _{n \rightarrow \infty} \bar{\kappa}\left(e_{1}, \frac{\alpha}{n} S_{U}^{n}, \kappa_{\text {init }}\right)=\bar{\kappa}^{\infty}\left(\left(\left[U^{i}\right]_{1}\right)_{i=1, \ldots, \lambda}, \kappa_{\text {init }}\right)
$$

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

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

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

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

$$
\begin{aligned}
n \ln \| e_{1}+ & \bar{\kappa}_{n}\left(e_{1}, \alpha / n \sum_{i=1}^{\mu} w_{i} U_{n}^{i}\right) \frac{\alpha}{n} \sum_{i=1}^{n} w_{i} U_{n}^{i} \|^{2} \\
& =-8 \ln \left[\left(1-\frac{\left[\sum_{i=1}^{\mu} w_{i} U_{n}^{i}\right]_{1}^{2}}{\left\|\sum_{i=1}^{\mu} w_{i} U_{n}^{i}\right\|^{2}}\right)^{-n / 8}\right] \mathbf{1}_{\left[\sum_{i=1}^{\mu} w_{i} U_{n}^{i}\right]_{1}<0}
\end{aligned}
$$

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

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

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

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

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

$$
\begin{aligned}
& \left\|e_{1}+\bar{\kappa}_{n}\left(e_{1}, \alpha / n \sum_{i=1}^{\mu} w_{i} U_{n}^{i}\right) \alpha / n \sum_{i=1}^{n} w_{i} U_{n}^{i}\right\|^{2} \\
& \geqslant\left\|e_{1}+\bar{\kappa}_{\mathrm{PLS}}\left(e_{1}, \alpha / n \sum_{i=1}^{\mu} w_{i} U_{n}^{i}\right) \alpha / n \sum_{i=1}^{n} w_{i} U_{n}^{i}\right\|^{2}
\end{aligned}
$$

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

$$
\begin{align*}
& n \ln ^{-}\left\|e_{1}+\bar{\kappa}_{n}\left(e_{1}, \alpha / n \sum_{i=1}^{\mu} w_{i} U_{n}^{i}\right) \alpha / n \sum_{i=1}^{n} w_{i} U_{n}^{i}\right\|^{2} \\
& \quad \leqslant n \ln ^{-}\left\|e_{1}+\bar{\kappa}_{\mathrm{PLS}}\left(e_{1}, \alpha / n \sum_{i=1}^{\mu} w_{i} U_{n}^{i}\right) \alpha / n \sum_{i=1}^{n} w_{i} U_{n}^{i}\right\|^{2} \tag{1}
\end{align*}
$$

Note that, as $\left\|e_{1}+\bar{\kappa}_{\mathrm{PLS}}\left(e_{1}, \alpha / n \sum_{i=1}^{\mu} w_{i} U_{n}^{i}\right) \alpha / n \sum_{i=1}^{n} w_{i} U_{n}^{i}\right\| \leqslant \| e_{1}+$ $0 \|=1$, then $\ln \left\|e_{1}+\bar{\kappa}_{\text {PLS }}\left(e_{1}, \alpha / n \sum_{i=1}^{\mu} w_{i} U_{n}^{i}\right) \alpha / n \sum_{i=1}^{n} w_{i} U_{n}^{i}\right\| \leqslant$ $\left\|e_{1}+0\right\| \leqslant 0$, thus is equal in absolute value to its negative part.

However, as

$$
\left(n \ln ^{-}\left\|e_{1}+\bar{\kappa}_{\mathrm{PLS}}\left(e_{1}, \alpha / n \sum_{i=1}^{\mu} w_{i} U_{n}^{i}\right) \alpha / n \sum_{i=1}^{n} w_{i} U_{n}^{i}\right\|^{2}\right)_{n \in \mathbb{N}^{*}}
$$

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

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

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

$$
\ln ^{+}\left\|e_{1}+\kappa x\right\| \leqslant \ln ^{+}\left\|e_{1}+\kappa^{M} x\right\|
$$

Indeed, the above equation is equivalent to

$$
\left\|e_{1}+\kappa x\right\|^{2} \mathbf{1}_{\left\|e_{1}+\kappa x\right\| \geq 1} \leqslant\left\|e_{1}+\kappa^{M} x\right\|^{2} \mathbf{1}_{\left\|e_{1}+\kappa^{M} x\right\| \geq 1}
$$

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

For the positive part, we have then

$$
\begin{aligned}
n \ln ^{+} \| e_{1}+\bar{\kappa}_{n}\left(e_{1}, \frac{\alpha}{n} \sum_{i=1}^{\mu} w_{i} U_{n}^{i}\right) & \frac{\alpha}{n} \sum_{i=1}^{\mu} w_{i} U_{n}^{i} \|^{2} \\
& \leqslant n \ln ^{+}\left\|e_{1}+\frac{\kappa^{M} \alpha}{n} \sum_{i=1}^{\mu} w_{i} U_{n}^{i}\right\|^{2}
\end{aligned}
$$

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

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

is uniformly integrable.

## REFERENCES

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[^0]
[^0]:    ${ }^{1}$ replace $\sigma^{*}$ by $\alpha \kappa{ }^{M}$ in the proof of Proposition 4 of [2]

