

Overcoming the curse of dimensionality: from nonlinear Monte Carlo to the training of deep neural networks

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ANNs: For every $d \in \mathbb{N}$, $x = (x_1, \dots, x_d) \in \mathbb{R}^d$ let
 $\mathcal{A}(x) = (\max\{x_1, 0\}, \dots, \max\{x_d, 0\})$,
let $\mathbf{N} = \cup_{L \in \mathbb{N}} \cup_{\ell_0, \dots, \ell_L \in \mathbb{N}} (\times_{n=1}^L (\mathbb{R}^{\ell_n \times \ell_{n-1}} \times \mathbb{R}^{\ell_n}))$,

and $\forall L \in \mathbb{N}, \ell_0, \dots, \ell_L \in \mathbb{N}, U = ((W_1, B_1), \dots, (W_L, B_L)) \in \times_{n=1}^L (\mathbb{R}^{\ell_n \times \ell_{n-1}} \times \mathbb{R}^{\ell_n})$

let $\mathcal{R}_U: \mathbb{R}^{\ell_0} \rightarrow \mathbb{R}^{\ell_L}$ and $\mathcal{P}_U \in \mathbb{N}$ satisfy for all $x_0 \in \mathbb{R}^{\ell_0}, \dots, x_L \in \mathbb{R}^{\ell_L}$ with

$\forall n \in \{1, \dots, L\}: x_n = \mathcal{A}(W_n x_{n-1} + B_n)$ that

$$\mathcal{R}_U(x_0) = W_L x_{L-1} + B_L, \quad \mathcal{P}_U = \sum_{n=1}^L (\ell_n \ell_{n-1} + \ell_n).$$

Theorem (Hutzenthaler-J-Kruse-Nguyen 2020 PDEA; Ackermann.-Kruse-Kuck.-J-Padgett 2024)

Let $T, \kappa > 0$, let $f: \mathbb{R} \rightarrow \mathbb{R}$ be Lipschitz, $\forall d \in \mathbb{N}$ let $g_d \in C^1(\mathbb{R}^d, \mathbb{R})$ and
 $u_d: [0, T] \times \mathbb{R}^d \rightarrow \mathbb{R}$ be an at most poly. grow. solution of

$$\frac{\partial u_d}{\partial t} = \Delta_x u_d + f(u_d), \quad u_d(0, \cdot) = g_d$$

with $|g_d(x)| + \|(\nabla g_d)(x)\| \leq \kappa d^\kappa (1 + \|x\|^\kappa)$, and assume $\forall d \in \mathbb{N}, \varepsilon \in (0, 1]$:

$\exists G \in \mathbf{N}: \forall x \in \mathbb{R}^d: |g_d(x) - \mathcal{R}_G(x)| \leq \varepsilon \kappa d^\kappa (1 + \|x\|^\kappa)$ and $\mathcal{P}_G \leq \kappa d^\kappa \varepsilon^{-\kappa}$.

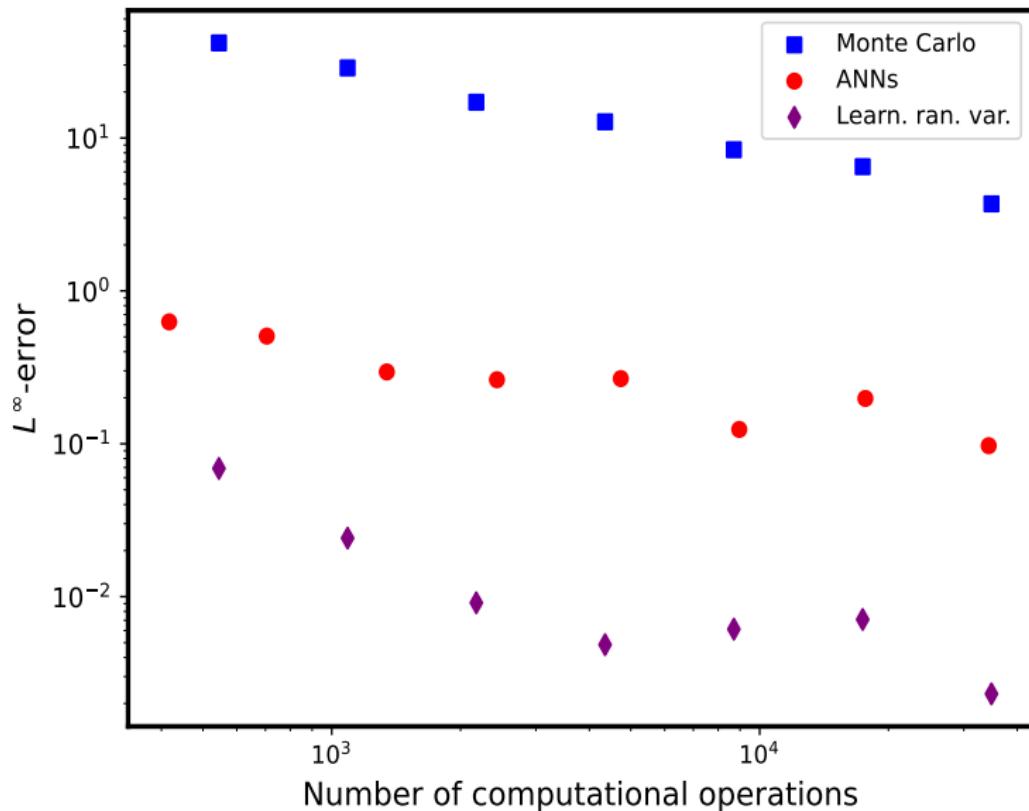
Then $\exists c > 0: \forall d \in \mathbb{N}, \varepsilon \in (0, 1]: \exists U \in \mathbf{N}:$

$$\int_{[0, T] \times [0, 1]^d} |u_d(y) - \mathcal{R}_U(y)| dy \leq \varepsilon \quad \text{and} \quad \mathcal{P}_U \leq c d^c \varepsilon^{-c}.$$

Hutzenthaler-J-Kruse-Nguyen-von Wurstemberger 2020 PRSA,

Gonon-Graeber-J 2023, Becker-J-Müller-von Wurstemberger 2023 Math. Finance

$$x \in [90, 110], \quad T \in [\frac{1}{100}, 1], \quad r \in [-\frac{1}{10}, \frac{1}{10}], \quad \sigma \in [\frac{1}{100}, \frac{1}{2}], \quad K \in [90, 110]$$



(addition, subtraction, multiplication, division, exponential, GeLU)

Let $d \in \mathbb{N}$, $a \in \mathbb{R}$, $b > a$, let $(\Omega, \mathcal{F}, \mathbb{P})$ be prob. sp., for every $m, n \in \mathbb{N}_0$ let $X_n^m: \Omega \rightarrow [a, b]^d$ and $Y_n^m: \Omega \rightarrow \mathbb{R}$ be RVs, assume for all $i \in \mathbb{N}, j \in \mathbb{N} \setminus \{i\}$ that $\mathbb{P}(X_0^i = X_0^j) = 0$, for every $k \in \mathbb{N}_0$ let $\mathfrak{d}_k, L_k \in \mathbb{N} \setminus \{1\}$, $\mathbf{l}_k = (\mathbf{l}_k^0, \dots, \mathbf{l}_k^{L_k}) \in \mathbb{N}^{L_k+1}$ satisfy $\mathbf{l}_k^0 = d$, $\mathbf{l}_k^{L_k} = 1$, and $\mathfrak{d}_k = \sum_{i=1}^L \mathbf{l}_k^i (\mathbf{l}_k^{i-1} + 1)$, for every $k, n \in \mathbb{N}_0$, $\theta \in \mathbb{R}^{\mathfrak{d}_k}$ let $\mathcal{N}_k^{\nu, \theta} = (\mathcal{N}_{k,1}^{\nu, \theta}, \dots, \mathcal{N}_{k,L_k}^{\nu, \theta}): \mathbb{R}^d \rightarrow \mathbb{R}^{\mathbf{l}_k^\nu}$, $\nu \in \{0, 1, \dots, L_k\}$, satisfy for all $x \in \mathbb{R}^d$, $i \in \{1, \dots, \mathbf{l}_k^{\nu+1}\}$ that

$$\begin{aligned} \mathcal{N}_{k,i}^{\nu+1, \theta}(x) &= \theta_{\mathbf{l}_k^{\nu+1} \mathbf{l}_k^\nu + i + \sum_{h=1}^\nu \mathbf{l}_k^h (\mathbf{l}_k^{h-1} + 1)} \\ &+ \sum_{j=1}^{\mathbf{l}_k^\nu} \theta_{(i-1) \mathbf{l}_k^\nu + j + \sum_{h=1}^\nu \mathbf{l}_k^h (\mathbf{l}_k^{h-1} + 1)} (x_j \mathbb{1}_{\{0\}}(\nu) + \max\{\mathcal{N}_{k,j}^{\nu, \theta}(x), 0\} \mathbb{1}_{\mathbb{N}}(\nu)), \end{aligned}$$

for every $k, n \in \mathbb{N}_0$ let $M_n^k \in \mathbb{N}$, $\gamma_n^k \in \mathbb{R}$, let $\mathcal{R}_n^k: \mathbb{R}^{\mathfrak{d}_k} \times \Omega \rightarrow \mathbb{R}$ satisfy

$$\mathcal{R}_n^k(\theta) = \frac{1}{M_n^k} \left[\sum_{m=1}^{M_n^k} |\mathcal{N}_k^{L_k, \theta}(X_n^m) - Y_n^m|^2 \right],$$

let $\mathfrak{G}_n^k: \mathbb{R}^{\mathfrak{d}_k} \times \Omega \rightarrow \mathbb{R}^{\mathfrak{d}_k}$ be generalized gradient of \mathcal{R}_n^k , and let $\Theta_n^k: \Omega \rightarrow \mathbb{R}^{\mathfrak{d}_k}$ be RV, assume $\forall k, n \in \mathbb{N}: \Theta_n^k = \Theta_{n-1}^k - \gamma_n^k \mathfrak{G}_n^k(\Theta_{n-1}^k)$, $\liminf_{k \rightarrow \infty} \mathbf{l}_k^1 = \infty$, and $\liminf_{k \rightarrow \infty} \mathbb{P}(\inf_{\theta \in \mathbb{R}^{\mathfrak{d}_k}} \mathcal{R}_0^k(\theta) > 0) = 1$, let $(c_k)_{k \in \mathbb{N}} \subseteq (0, \infty)$ satisfy for all $k \in \mathbb{N}$ that $c_k \Theta_0^k$ is standard normal. Then

$$\liminf_{k \rightarrow \infty} \mathbb{P} \left(\inf_{n \in \mathbb{N}_0} \mathcal{R}_0^k(\Theta_n^k) > \inf_{\theta \in \mathbb{R}^{\mathfrak{d}_k}} \mathcal{R}_0^k(\theta) \right) = 1.$$

Let $(\Omega, \mathcal{F}, \mathbb{P})$ be prob. sp., let $d \in \mathbb{N}$, $a \in \mathbb{R}$, $b > a$, let $X_{n,m}: \Omega \rightarrow [a, b]^d$, $(n, m) \in \mathbb{N}^2$, be i.i.d. RVs, let $\mathfrak{l}: \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$ satisfy for all $\theta, x \in \mathbb{R}^d$ that

$$\mathfrak{l}(\theta, x) = \|\theta - x\|^2,$$

let $\xi \in \mathbb{R}^d$, $p, \varepsilon, \alpha \in (0, \infty)$, $\beta \in (\alpha^2, 1)$, $(\gamma_n)_{n \in \mathbb{N}} \subseteq (0, \infty)$ satisfy

$\limsup_{n \rightarrow \infty} (\gamma_n + \gamma_n^{-2} |\gamma_n - \gamma_{n+1}|) = 0$, for every $M \in \mathbb{N}$ let $\mathcal{M}^M: \mathbb{N}_0 \times \Omega \rightarrow \mathbb{R}^d$,

$\mathbb{M}^M: \mathbb{N}_0 \times \Omega \rightarrow \mathbb{R}^d$, and $\Theta^M: \mathbb{N}_0 \times \Omega \rightarrow \mathbb{R}^d$ be stoch. proc. satisfying for all

$n \in \mathbb{N}$, $i \in \{1, \dots, d\}$ that

$$\mathcal{M}_0^M = 0, \quad \mathcal{M}_n^M = \alpha \mathcal{M}_{n-1}^M + \frac{(1-\alpha)}{M} \left[\sum_{m=1}^M (\nabla_\theta \mathfrak{l})(\Theta_{n-1}^M, X_{n,m}) \right],$$

$$\mathbb{M}_0^{M,i} = 0, \quad \mathbb{M}_n^{M,i} = \beta \mathbb{M}_{n-1}^{M,i} + \frac{(1-\beta)}{M^2} \left[\sum_{m=1}^M \left(\frac{\partial}{\partial \theta_i} \mathfrak{l} \right)(\Theta_{n-1}^M, X_{n,m}) \right]^2,$$

$$\Theta_0^M = \xi, \quad \Theta_n^{M,i} = \Theta_{n-1}^{M,i} - \gamma_n [\varepsilon + [(1-\beta)^{-1} \mathbb{M}_n^{M,i}]^{1/2}]^{-1} \mathcal{M}_n^{M,i},$$

Then there exist $c > 0$, $(\vartheta_M)_{M \in \mathbb{N}} \subseteq \mathbb{R}^d$ such that for all $n \in \mathbb{N}$, $M \in \mathbb{N} \cap [c, \infty)$:

- $\mathbb{P}(\limsup_{N \rightarrow \infty} \|\Theta_N^M - \vartheta_M\| = 0) = 1$,

- $(\mathbb{E}[\|\Theta_n^M - \vartheta_M\|^p])^{1/p} \leq c \sqrt{\gamma_n}$, and

- $\|\vartheta_M - \mathbb{E}[X_{1,1}]\| \leq c M^{-1}$.

Many thanks for your attention!



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