

Universal Average-Case Optimality of Polyak Momentum



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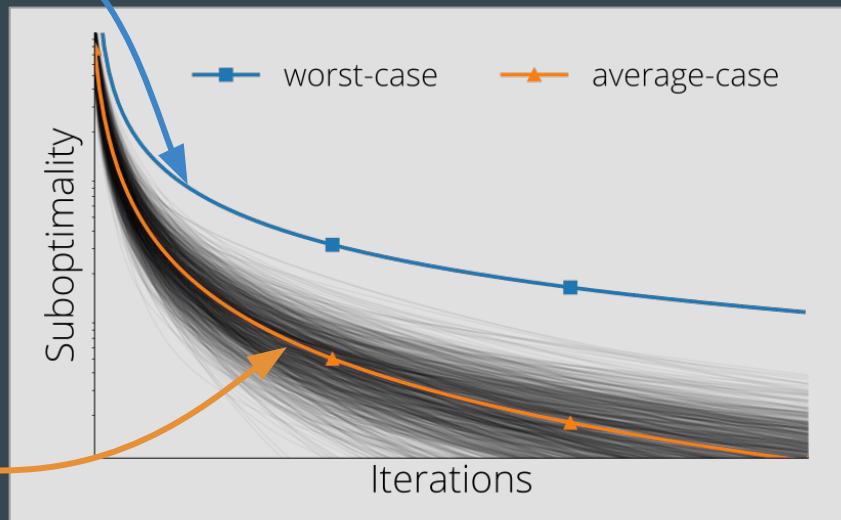
Worst Case V.S. Average Case

Worst case

- ✓ Complexity bounds for any input.
- ✗ Not representative of typical runtime.

Average case

- ✓ Representative of the typical behaviour.
- ??? Complexity averaged over all problem instances.



Optimal Average Case Methods

Best method: minimizes P_t in

$$\overbrace{R^2}^{\text{initialization}} \int_{\mathbb{R}} \underbrace{P_t^2}_{\text{algorithm}} \overbrace{d\mu}^{\text{problem}}$$

Theorem Pedregosa, Scieur (ICML 2020)

Any optimal average-case method has the form

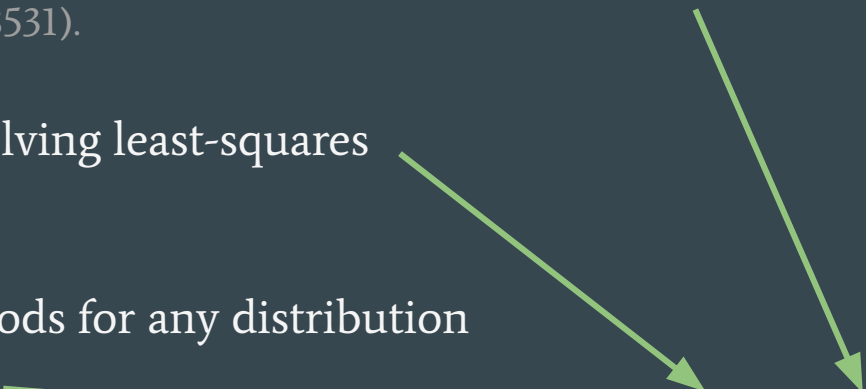
$$\mathbf{x}_t = \mathbf{x}_{t-1} + (1 - a_t)(\mathbf{x}_{t-2} - \mathbf{x}_{t-1}) + b_t \nabla f(\mathbf{x}_{t-1})$$

Momentum

Gradient step-size

Optimal Average Case Methods: Applications

- Neural networks share similar training dynamics of a quadratic problem.
(Jacot et al. 2018, Novak et al. 2018, Arora et al. 2019, Chizat and Bach, 2019)
- Design of accelerated gossip algorithms: optimal method w.r.t. Jacobi measure
Berthier, Bach, Gaillard, (arxiv 1805.08531).
- Random matrix sketching for solving least-squares
Lacotte, Pilanci (ICML 2020)
- Possible to design optimal methods for any distribution
Pedregosa, Scieur (ICML 2020)



Noticed some regularity when
#iterations goes to infinity

Conjecture

**All average-case optimal methods
converge to Polyak momentum**

(in $\#$ iterations, whatever the expected spectral density)

Asymptotic Rate of Optimal Methods (Scieur & Pedregosa, ICML 2020)

Assume we use an **average-case optimal method** w.r.t. the density function μ , **strictly positive** on the interval $[\ell, L]$. Then, *for all such densities μ ,*

$$\lim_{t \rightarrow \infty} \sqrt[t]{\mathbb{E} \left[\frac{\|\mathbf{x}_t - \mathbf{x}^*\|^2}{\|\mathbf{x}_0 - \mathbf{x}^*\|^2} \right]} = \left(\frac{\sqrt{L} - \sqrt{\ell}}{\sqrt{L} + \sqrt{\ell}} \right)^2 .$$

Expectation over
problem instances

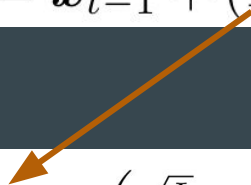
Relative rate of
convergence

Rate of Polyak
Momentum


Asymptotic Recurrence (Scieur & Pedregosa, ICML 2020)

Assume we use an **average-case optimal method** w.r.t. the density function μ , **strictly positive** on the interval $[\ell, L]$. Then, *for all such densities μ ,*

$$\mathbf{x}_t = \mathbf{x}_{t-1} + (1 - a_t)(\mathbf{x}_{t-2} - \mathbf{x}_{t-1}) + b_t \nabla f(\mathbf{x}_{t-1})$$

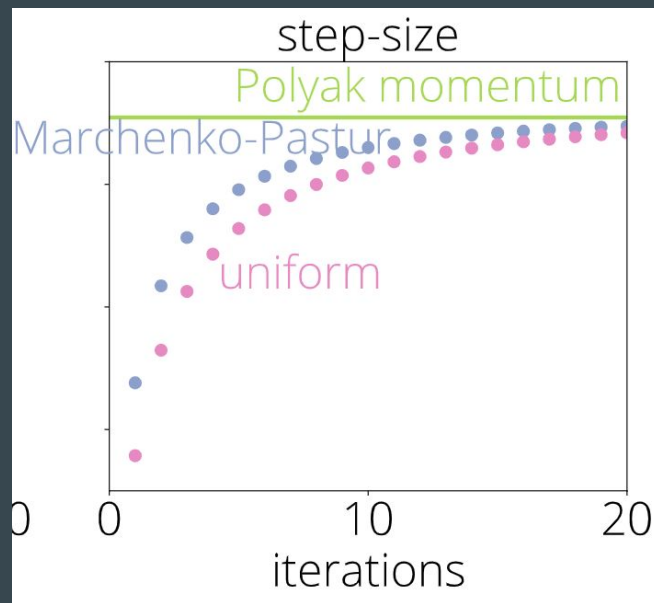
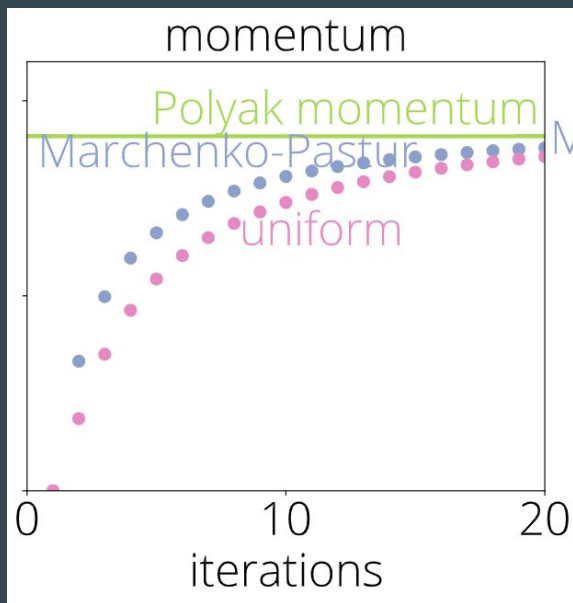
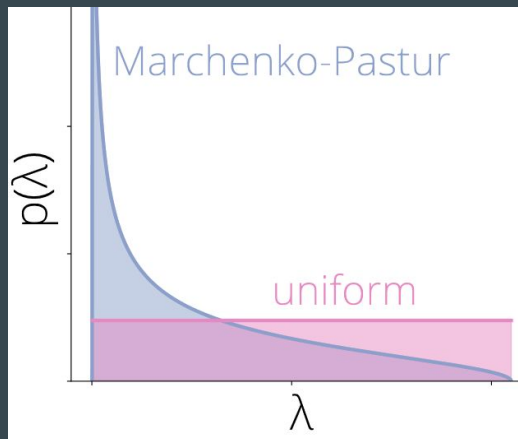

$$\lim_{t \rightarrow \infty} (1 - a_t) = - \left(\frac{\sqrt{L} - \sqrt{\ell}}{\sqrt{L} + \sqrt{\ell}} \right)^2$$

Polyak's optimal momentum


$$\lim_{t \rightarrow \infty} b_t = - \left(\frac{2}{\sqrt{L} + \sqrt{\ell}} \right)^2$$

Polyak's optimal step-size

Numerical evidences



Take-home message

**Polyak momentum is *provably*
always a good choice.**

- Easier to design than optimal method
- Strong theory explaining its good empirical performance
- Possible loss: constant number of iterations