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Invited Paper


# Machine learning for quantitative finance: fast derivative pricing, hedging and fitting

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## Abstract

In this paper, we show how we can deploy machine learning techniques in the context of traditional quant problems. We illustrate that for many classical problems, we can arrive at speed-ups of several orders of magnitude by deploying machine learning

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## Disclosure statement

No potential conflict of interest was reported by the authors.

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## Notes

† Matrix inversion is often implemented via a Cholesky decomposition (Benoît [1924](#), Rasmussen and Williams [2006](#)), which is more stable than actually inverting the matrix. For small matrices, i.e. small values of  $n$ , ordinary matrix inversion can be performed. For the results in this paper we used the Matlab functions `fitrgp` and `predict`. However if the dimension increases, special techniques need to be deployed. We mention LU-factorization and blockwise Cholesky decomposition, which aim at solving traditional memory problems that one encounters when inverting large matrices. For future work we will employ Cholesky and blockwise Cholesky routines to handle problems with many more data points.

†  $\kappa$  = rate of mean reversion,  $\rho$  = correlation stock - vol,  $\theta$  = vol of vol,  $\eta$  = long run variance,  $\sigma^2$  = initial variance.

† For each  $\sigma$  we construct the 100



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