

Title	DOI	Notes
<i>Best-Practice DFT Protocols for Basic Molecular Computational Chemistry</i>	https://doi.org/10.1002/anie.202205735	Perspective on best-practice protocols and guidance in the choice of robust quantum-chemical methods to model experiments as close a possible
<i>A binding mode hypothesis for prothioconazole binding to CYP51 derived from first principles quantum chemistry</i>	https://doi.org/10.1007/s10822-020-00331-z	Binding Mode Hypotheses for Prothioconazole
<i>Flupyradifurone: a brief profile of a new butenolide insecticide</i>	https://doi.org/10.1002/ps.3932	Discovery of Butenolides
<i>Bayer's in silico ADMET platform: a journey of machine learning over the past two decades</i>	https://doi.org/10.1016/j.drudis.2020.07.001	Overview of the Machine Learning Model Platform at Bayer Pharma with details on data curation, descriptors, and algorithms.
<i>MELLODDY: Cross-pharma Federated Learning at Unprecedented Scale Unlocks Benefits in QSAR without Compromising Proprietary Information</i>	https://doi.org/10.1021/acs.jcim.3c00799	Main outcome of the IMI MELLODDY collaboration on Machine Learning across multiple partners
<i>MELLODDY GitHub Repository</i>	https://github.com/melloddy	Open-source ML tool developed within the MELLODDY collaboration
<i>Site of Metabolism Prediction Based on ab initio Derived Atom Representations</i>	https://doi.org/10.1002/cmdc.201700097	Example of model improvements from QM-informed descriptors
<i>!AIQU Announcement</i>	https://www.linkedin.com/posts/tobias-morawietz_machinelearning-neuralnetworks-artificialintelligence-activity-6891744599526703104-t19W?utm_source=share&utm_medium=member_android	Public announcement of the !AIQU collaboration on combining QM and AI for accelerated molecular design
<i>Machine learning-accelerated quantum mechanics-based atomistic simulations for industrial applications</i>	https://doi.org/10.1007/s10822-020-00346-6	Overview of ML-accelerated QM simulations applied to industrial applications
<i>An implementation of artificial neural-network potentials for atomistic materials simulations: Performance for TiO2</i>	https://doi.org/10.1016/j.commatsci.2015.11.047	Open-source code for training Machine Learning models for atomistic simulations
<i>ænet-PyTorch: A GPU-supported implementation for machine learning atomic potentials training</i>	https://doi.org/10.1063/5.0146803	Extended code for efficient training on GPUs