

Machine Learning: The Way for Improving Law Application

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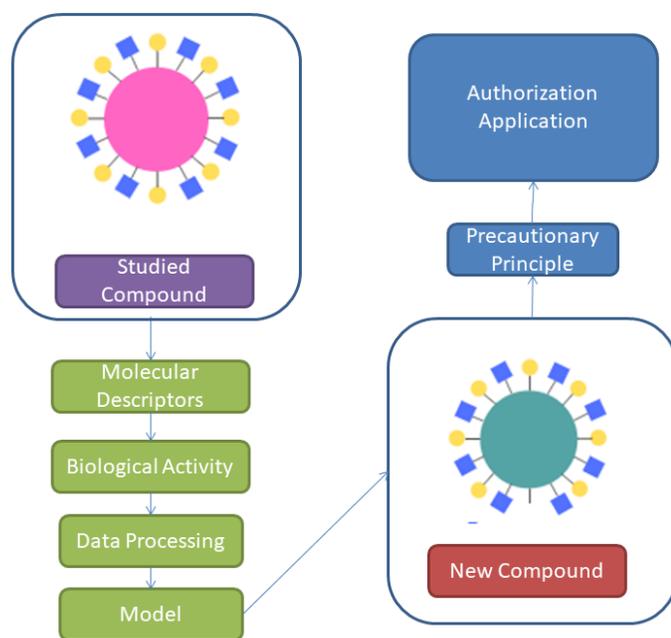
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Abstract

The study was carried out to show how the Precautionary Principle is better applied with Machine Learning techniques. The development of Machine Learning techniques and the wider application for different disciplines, such as Biotechnology, Nanotechnology, and Science of Materials, has been increasing.¹⁻¹³ Modelling and simulation techniques should be promoted for regulation assessment, taking into consideration the uncertainty of new compounds product of biotechnology and nanotechnology applications.



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