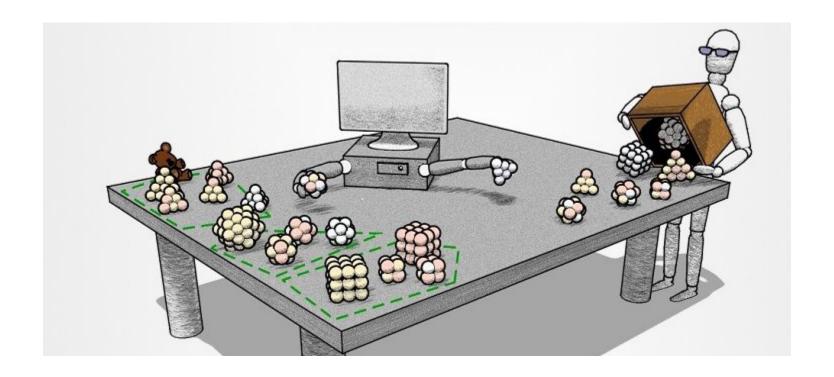
MACHINE-LEARNING IN **CHEMISTRY**



Filippo Federici

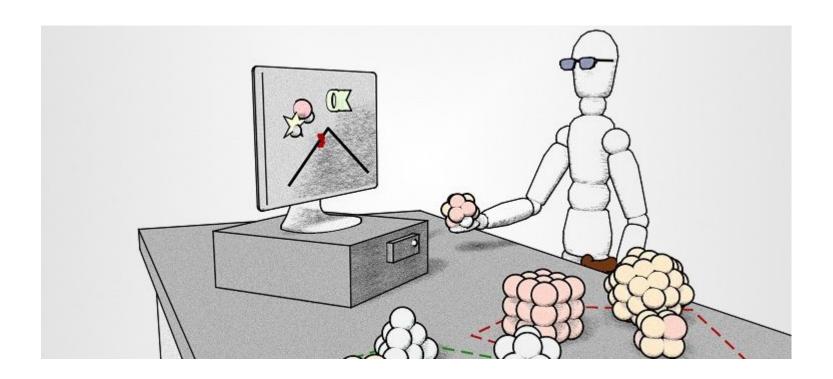


UNSUPERVISED LEARNING



- finds similarities in complex data records
- does not require knowledge of properties/outputs, only descriptors/inputs
- sensitive to the similarity measure
- requires the user to know how many classes to expect
- useful to reduce data dimensionality

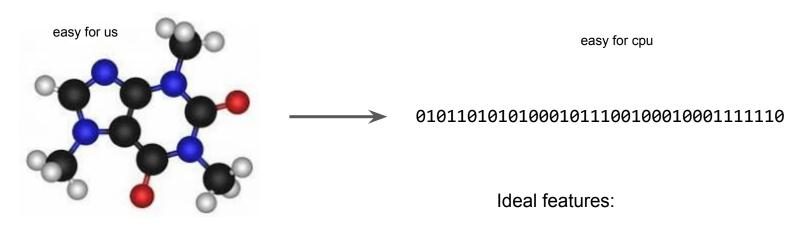
SUPERVISED LEARNING



- $\bullet \quad \text{learns input} \to \text{output relation from examples}$
- training data is the limit
- useful for fast screening and classification

DESCRIPTORS FOR CHEMISTRY

ML methods need a computer-friendly way to input the atomistic system:



Issues for ML:

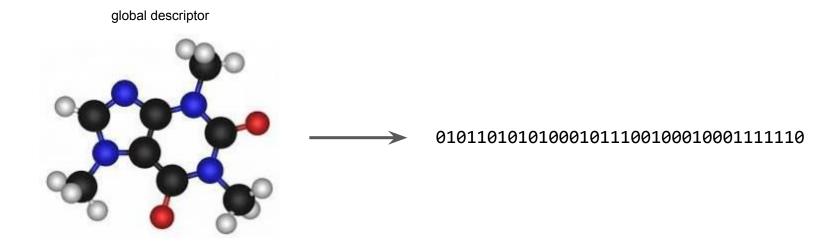
- arbitrary size
- arbitrary order

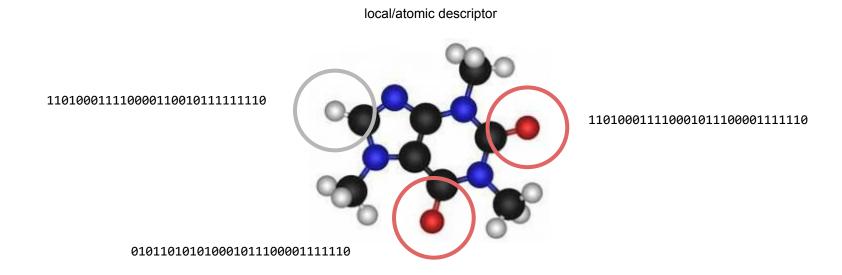
- general
- compact
- unique
- invariant *
- smooth
- fast

^{*} invariants are determined by the physics of the quantity to predict from the descriptor!

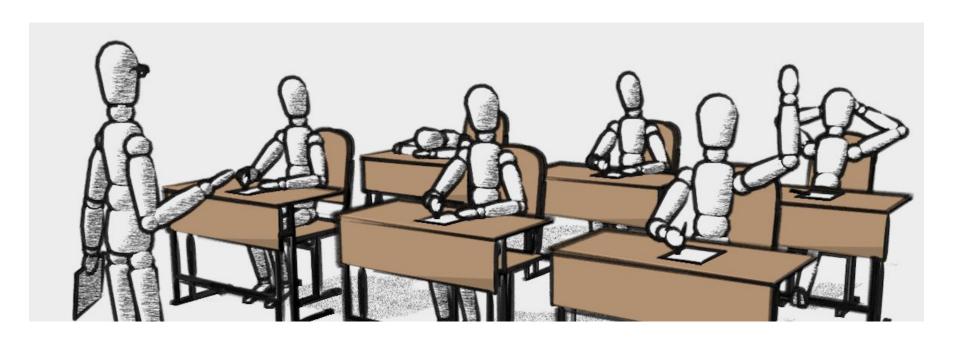
DESCRIPTORS FOR CHEMISTRY

ML methods need a computer-friendly way to input the atomistic system:



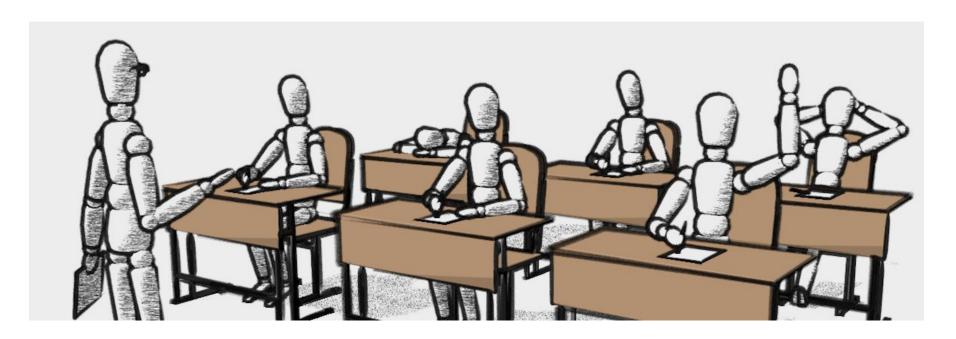


DESCRIPTORS FOR CHEMISTRY



- 1. ACSF.ipynb
- 2. SOAP.ipynb
- 3. MBTR.ipynb
- 4. LMBTR.ipynb

UNSUPERVISED ML METHODS



- 1. PCA.ipynb
- 2. tSNE.ipynb

SUPERVISED ML METHODS

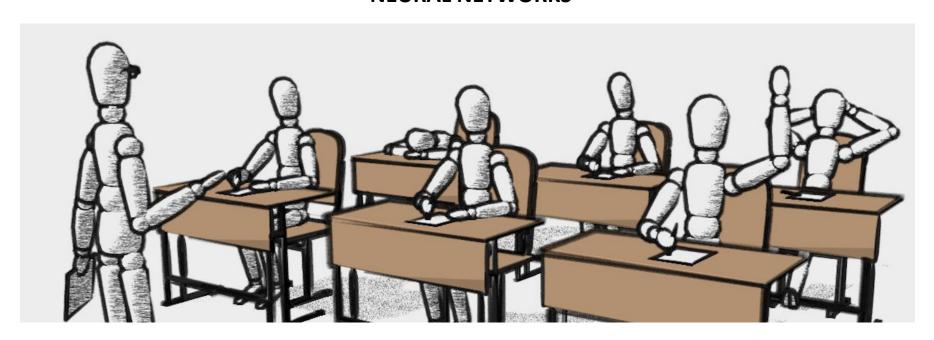
KERNEL RIDGE REGRESSION



KRR - TotalEnergy.ipynb

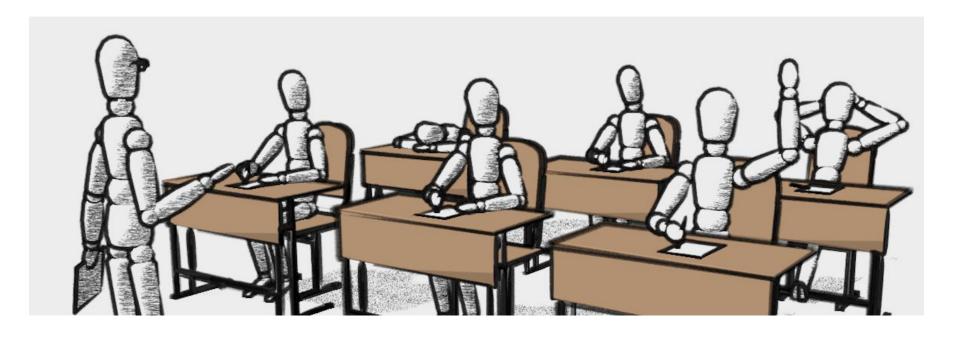
SUPERVISED ML METHODS

NEURAL NETWORKS



- 1. NeuralNetwork Intro.ipynb
- 2. ACSF-Dimer.ipynb
- 3. NeuralNetwork TotalEnergy.ipynb
- 4. NeuralNetwork AtomicCharges.ipynb
- 5. GenericAlgorithm.ipynb

TAKEHOME



- all notebooks and data is in github: https://github.com/fullmetalfelix/ML-CSC-tutorial
 - o notebooks require Jupyter python module
 - o data in numpy array form
- useful goodies:
 - describe package: https://github.com/SINGROUP/describe
 - python package for creating machine learning descriptors for atomistic systems



