SOAPLite Documentation

December 10, 2018

Abstract

SOAP is a state-of-the-art local chemical descriptor that has achieved amazing accuracies in machine learning in chemistry [1,2,3,4]. SoapLite is a fast and lightweight SOAP-spectra [1,2] calculator that gives local chemical descriptors of 1) arbitrary positions in a chemical structure, or 2) every atom position in a chemical structure, both with periodic and non-periodic boundary conditions. SOAPLite was designed for ease-of-use, portability and computational performance in mind, with the potential to perform real-time SOAP analysis. Computation of the kernel is not included, and must be done independently such as with scikit-learn, by implementing from scratch, or by other ready-made SOAP implementations such as with glosim [5].

1 Derivation of SOAP Spectrum

The core algorithm is based on the papers in the reference [1,2]. In SOAPLite, the radial basis $g_{nl}^{(l)}$ and atomic density ρ are

$$g_{nl}^{(l)}(r) = \sum_{k=1}^{N_k} \beta_{nk}^{(l)} r^l e^{-\alpha_{kl} r^2}, \tag{1}$$

$$\rho_{N_A}^{\mu}(\mathbf{r}) = \sum_{i=1}^{N_A} e^{-(\mathbf{r} - \mathbf{r}_i^{\mu})^2} = \sum_{i=1}^{N_A} e^{-(x^2 + y^2 + z^2 - 2(xx_i^{\mu} + yy_i^{\mu} + zz_i^{\mu}))} e^{-(x_i^{\mu^2} + y_i^{\mu^2} + z_i^{\mu^2})}.$$
(2)

where μ is the type of the atom, N_A is the number of atoms from the center (the desired local position \mathbf{x} of SOAP spectrum translated to the origin), x, y and z are the position in the Cartesian coordinates, α, β give the set of orthonormal radial basis functions and N_k is the number of radial basis functions. The SOAP power spectrum at a certain position \mathbf{x} is defined as

$$\mathbf{P} = P_{nn'l}^{\mu\nu}(\mathbf{x}) = \sum_{m} c_{nlm}^{\mu}(\mathbf{x}) c_{n'lm}^{*\nu}(\mathbf{x})$$
(3)

where

$$c_{nlm}^{\mu} = \int dV g_{nl}^{(l)}(r) \rho_{N_A}(\mathbf{r}) Y_{lm}(\theta, \phi)$$
(4)

with different atom types μ, ν and by summing m, the spectra is guaranteed to be rotationally invariant. Y_{lm} are the spherical harmonics.

To find $\beta_{nk}^{(l)}$, we need the analytical formulas

$$\int_0^\infty dr \, r^{2(l+1)} e^{-cr^2} = \frac{1}{2} \Gamma \left(l + \frac{3}{2} \right) c^{-l-3/2} \tag{5}$$

$$\int_0^x dr \, r^{2(l+1)} e^{-cr^2} = \frac{x^{2l-1}}{2c^2} (cx^2)^{1/2-l} \left(\Gamma\left(l + \frac{3}{2}\right) - \Gamma\left(l + \frac{3}{2}, cx^2\right) \right) \tag{6}$$

where Γ is the Gamma function, for every l, we can build a matrix of $N_k \times N_k$ by integrating over several different α 's, we get

$$G_{nn'}^{(l)} = \int_0^\infty dr \, r^2 \, r^l e^{-\alpha_{nl} r^2} e^{-\alpha_{n'l} r^2} \tag{7}$$

Finally, by inverting $G_{nn'}$ and take the matrix square-root, we systematically get the β

$$\beta^{(l)} = G_{(l)}^{-1/2} \tag{8}$$

In SOAPLite, the α_n 's are selected so that it damps at every point from +1 Å to $r_{\text{soft_cut}}$, spaced evenly, with a hard cutoff to avoid unnecessary integrations that will give zero: $r_{\text{hard_cut}} = r_{\text{soft_cut}} + 5\text{Å}$.

2 Code Description

2.1 Navigation

2.1.1 Main Directory

./soapPy.py is the main program that wraps the source codes. ./genBasis.py is the basis generator for the soap, which follows the procedure shown in the Derivation of SOAP Spectrum section. ./batchSoapPy.py is used in case of getting SOAP spectra from a trajectory file. Structs contains structures (tests/*.xyz and tests/*.pdb) and positions in space (tests/*.dat) one can test. tests contains test python executables with periodic and non-periodic cases. src contains the source files written in c.

2.1.2 Source Directory

The naming convention in ./src is as follows: ./src/*Py* requires a main function or a wrapper (since they are precompiled libraries) and ./src/*Cro* implied that species crossover terms are edabled in $\mathbf{P}^{\mu\nu}$, so without crossover the spectrum will be $\mathbf{P}^{\mu\mu}$. The numbers on the ./src/*Cro* files indicate how

many species there are (for example, for 6 species, there will be 21 different combinations), which is branched in the python wrapper (./soapPy.py) depending on the number of species in the system.

2.2 Source Codes

Although the code seems overwhelming with numbers, they are just preset or precalculated numbers that are used to get $c_{nn'l}$ that are shown in the appendix.

2.2.1 Inputs

The inputs in the soap() in the source code are the pointer from the c wrapper in soapPy.py

```
double* c
```

Atom positions, and desired positions (Apos) for the power spectrum (Hpos)

```
double* Apos; double* Hpos;
```

 α 's, and β 's from the basis functions

```
double* alphas; double* betas;
```

numbers for each atom species as an array, $r_{\rm cut}$, numbers of atoms in the structure, number of species, number of basis functions, $L_{\rm max}$ and number of desired positions for the power spectrum.

```
int* typeNs ;double rCut; int totalAN;
int Nt; int Ns; int lMax; int Hs;
```

2.2.2 Embedded Arrays

Each vectors, matrices and tensors are stored respectively in a memory allocated array with a pointer, and the values are extracted or manipulated by shifting the positions and dimensions. The arrays are: The positions of the atoms in xyz-coordinate

```
double* x = malloc(sizeof(double)*totalAN);
double* y = malloc(sizeof(double)*totalAN);
double* z = malloc(sizeof(double)*totalAN);
```

 z^X where z is the z component of the atoms, and r^X where r is the distance from the reference position and X=2,3,4,5,6,7,8

```
double* zX = malloc(sizeof(double)*totalAN);
double* rX = malloc(sizeof(double)*totalAN);
```

Real and Imaginary part of x component and y component in the form $(x+iy)^X$

```
double* ReImX = malloc(2*sizeof(double)*totalAN);
```

precalculated exponents and preset coefficients where 96 was the minimum number coefficients required for up to $L_{\rm max}=9$ that isn't 1 in the paranthesis of the right hand side in the appendix's formulas.

```
double* exes = malloc (sizeof(double)*totalAN);
double* preCoef = malloc(96*sizeof(double)*totalAN);
```

precaculated $\alpha/(1+\alpha)$ and $\beta/(1+\alpha)$ (see Appendix)

```
double* b0a = malloc((lMax+1)*NsNs*sizeof(double));
double* a0a = malloc((lMax+1)*Ns*sizeof(double));
```

and finally c_{nlm} and $P_{nn'l}$

```
double* cnnd = malloc(100*Nt*Ns*Hs*sizeof(double));
double* soapMat =
    malloc(Hs*((Ts*(Ts+2))/2)*Ns*Ns*(lMax+1)*sizeof(double));
```

For a non-crossover case, Ts * (Ts + 1)/2 is replaced with Ts. The detailed coefficients and formulas are in the appendix.

2.2.3 Functions and SOAP Output

Computing $(x+iy)^X$ for X=2,3,4,5,6,7,8. The reason for the many functions is to speed up the multiplications for example $(x+iy)^5=(x+iy)^3*(x+iy)^2$ where the right-hand side has been already precomputed, is faster than multiplying (x+iy) 5 times. Every odd array position is the imaginary part and the even position (including 0) is the real part. This makes the size of the ReImX arrays $2\times N_A$

```
double* getReIm2(); double* getReIm3();
void getMulReIm(); void getMulDouble()
```

and similarly, for r^X and z^X with X = 2, 3, 4, 5, 6, 7, 8

```
double* getRsZs()
```

To filter out the atoms faraway that would just give zeros for the integrations, Apos is rearranged by neighborhood and spits out the number of neighbors

```
int getFilteredPos()
```

the functions to precalculate the $\beta/(1+\alpha)$ and $\alpha/(1+\alpha)$ are

```
void getAlphaBeta()
```

and the following function gets c_{nml}

```
int getC()
```

finally, by summing $\sum_m c^\mu_{nml} c^{*\nu}_{n'ml}$ with species μ , we have the soapSpectra which is the output $P^{\mu\nu}_{nn'l}$

```
double* soap()
```

The size of **P** is $(H_s(N(N+1))/2\times((L+1)(T_s(T_s+1)/2))$, with the crossover and $(H_s(N(N+1))/2\times((L_{max}+1)Ts))$ without, where H_s are the number of positions, N is the number of basis functions L is L_{max} , and T_s is the number of species.

3 Tests and Examples

3.1 Symmetry Test

Here, translational and rotational invariance of the SOAP spectra is tested with a Au₄₀Cu₄₀ cluster and a SiH₄ molecule by running

```
python test_symmetry.py
```

in the tests/ directory. For $Au_{40}Cu_{40}$, the structure is rotated 45 degrees around x,y, and z-axis and translated in a few directions and the soap spectra is compared taking the maximum differences. For SiH_4 , the soap spectra of hydrogen are compared to each other.

3.2 examples

In the tests directory, there are files called example_non_periodic_SoapPy.py and example_periodic_SoapPy.py. The procedures are as follows:

- (1) Define structure
- (2) Define desired positions for SOAP spectra
- (3) Set radial basis functions
- (4) Get SOAP spectra (CAUTION: Make sure to use the same settings as the set basis function in (3))
 - (5) Save SOAP spectra in a text file

in 4), we can switch functions between if it is period, non periodic, locals or structures depending on the application. For locals, we need to define positions, while for structures, the positions will be assigned to each atoms in the structure.

The output is a matrix of (Number of defined positions \times Descriptor size). Thus, Each row corresponds to one string of soap spectrum. We can compare two soap spectra by taking two rows (within the same matrix or different matrices if the descriptor size is the same).

4 Possible Improvements and Speedups

Besides eleminating the transpose symmetry of the power spectra, here are possible improvements that can be done on the algorithm to speed of the execution.

- 1) at least in our system, the majority of the time is spent on computing the exponents. Using a less accurate but a faster exponential function might speed up the code significantly, such as by tabulating the exponents at relevant domains.
- 2) for every position, the code calculates the distance²'s to filter out faraway atoms that don't contribute to the integration. This is negligible compared to (1) unless the molecar structure is huge. This can be optimized by rearanging the order of the atoms in the input file before, or using other filtering schemes.
- 3) the radial basis function is produced in python (genBasis.py) that gives the α 's and β 's, however this is only to generalize the process, but not recomended if SOAPLite is embedded in another software for speed, and α 's and β 's should be tabulated instead.
- 4) RAM usage is not highly optimized, so if the structure doesn't contain atom numbers in the millions, this shouldn't be a problem, but if a massive structure is present, RAM usage and (2) is nessesary to be optimized.

If you have any questions, suggestions or comments please contanct: eiaki.morooka@aalto.fi

References

- [1] Albert P. Bartók, Risi Kondor and Gábor Csányi On representating chemical environments. https://arxiv.org/abs/1209.3140
- [2] Sandi De, Albert Bartók, Gábor Csányi and Michele Ceriotti Comparing molecules and solids across structural and alchemical space. https://arxiv.org/abs/1601.04077
- [3] Nongnuch Artrith, Alexander Urban and Gerbrand Ceder Efficient and Accurate Machine-Learning Interpolation of Atomic Energies in Compositions with Many Species. https://arxiv.org/abs/1706.06293
- [4] Albert P. Bartók, Sandip De, Carl Poelking, Noam Bernstein, James R. Kermode, Gábor Csányi, Michele Ceriotti *Machine learning unifies the modeling of materials and molecules* https://arxiv.org/pdf/1706.00179.pdf
- [5] https://github.com/cosmo-epfl/glosim

A Integrations

The following formulas are the analytical forms of the integrations for c_{nlm} .

$$c_{n00} = \sum_{k=1}^{3} \frac{\pi \beta_{nk}}{2(1+\alpha_{k0})^{\frac{3}{2}}} \sum_{i=1}^{N_A} e^{-\frac{\alpha_{k0}}{1+\alpha_{k0}}} r_i^2$$

$$\begin{split} c_{n10} &= \sum_{k=1}^{N_k} \frac{\sqrt{3}\pi\beta_{nk}}{2(1+\alpha_{k1})^{\frac{5}{2}}} \sum_{i=1}^{N_A} e^{-\frac{\alpha_{k1}}{1+\alpha_{k1}}} r_i^2_{z_i} \\ c_{n11} &= -\sum_{k=1}^{N_k} \frac{\sqrt{3}\pi\beta_{nk}}{2\sqrt{2}(1+\alpha_{k1})^{\frac{5}{2}}} \sum_{i=1}^{N_A} e^{-\frac{\alpha_{k1}}{1+\alpha_{k1}}} r_i^2_{i} (x_i+iy_i) \end{split}$$

$$\begin{split} c_{n20} &= \sum_{k=1}^{N_k} \frac{\sqrt{5}\pi\beta_{nk}}{4(1+\alpha_{k2})^{\frac{7}{2}}} \sum_{i=1}^{N_A} e^{-\frac{\alpha_{k2}}{1+\alpha_{k2}}r_i^2} (3z^2 - r^2) \\ c_{n21} &= -\sum_{k=1}^{N_k} \frac{\sqrt{15}\pi\beta_{nk}}{2\sqrt{2}(1+\alpha_{k2})^{\frac{7}{2}}} \sum_{i=1}^{N_A} e^{-\frac{\alpha_{k2}}{1+\alpha_{k2}}r_i^2} (x_i + iy_i) z_i \\ c_{n22} &= \sum_{k=1}^{N_k} \frac{\sqrt{15}\pi\beta_{nk}}{4\sqrt{2}(1+\alpha_{k2})^{\frac{7}{2}}} \sum_{i=1}^{N_A} e^{-\frac{\alpha_{k2}}{1+\alpha_{k2}}r_i^2} (x_i + iy_i)^2 \end{split}$$

$$\begin{split} c_{n30} &= \sum_{k=1}^{N_k} \frac{\sqrt{7}\pi\beta_{nk}}{4(1+\alpha_{k3})^{\frac{9}{2}}} \sum_{i=1}^{N_A} e^{-\frac{\alpha_{k3}}{1+\alpha_{k3}}} r_i^2 z_i (5z^2 - 3r^2) \\ c_{n31} &= -\sum_{k=1}^{N_k} \frac{\sqrt{21}\pi\beta_{nk}}{8(1+\alpha_{k3})^{\frac{9}{2}}} \sum_{i=1}^{N_A} e^{-\frac{\alpha_{k3}}{1+\alpha_{k3}}} r_i^2 (x_i + iy_i) (4z_i^2 - x_i^2 - y_i^2) \\ c_{n32} &= \sum_{k=1}^{N_k} \frac{\sqrt{105}\pi\beta_{nk}}{4\sqrt{2}(1+\alpha_{k3})^{\frac{9}{2}}} \sum_{i=1}^{N_A} e^{-\frac{\alpha_{k3}}{1+\alpha_{k3}}} r_i^2 (x_i + iy_i)^2 z_i \\ c_{n33} &= -\sum_{k=1}^{N_k} \frac{\sqrt{35}\pi\beta_{nk}}{8(1+\alpha_{n3})^{\frac{9}{2}}} \sum_{i=1}^{N_A} e^{-\frac{\alpha_{k3}}{1+\alpha_{k3}}} r_i^2 (x_i + iy_i)^3 \end{split}$$

$$\begin{split} c_{n40} &= \sum_{k=1}^{N_k} \frac{3\pi\beta_{nk}}{16(1+\alpha_{k4})^{\frac{11}{2}}} \sum_{i=1}^{N_A} e^{-\frac{\alpha_{k4}}{1+\alpha_{k4}}r_i^2} (35z_i^4 - 30z_i^2r_i^2 + 3r_i^4) \\ c_{n41} &= -\sum_{k=1}^{N_k} \frac{3\sqrt{5}\pi\beta_{nk}}{8(1+\alpha_{k4})^{\frac{11}{2}}} \sum_{i=1}^{N_A} e^{-\frac{\alpha_{k4}}{1+\alpha_{k4}}r_i^2} (x_i + iy_i)z_i (7z_i^2 - 3r_i^2) \\ c_{n42} &= \sum_{k=1}^{N_k} \frac{3\sqrt{5}\pi\beta_{nk}}{8\sqrt{2}(1+\alpha_{k4})^{\frac{11}{2}}} \sum_{i=1}^{N_A} e^{-\frac{\alpha_{k4}}{1+\alpha_{k4}}r_i^2} (x_i + iy_i)^2 (7z_i^2 - r_i^2) \\ c_{n43} &= -\sum_{k=1}^{N_k} \frac{3\sqrt{35}\pi\beta_{nk}}{8(1+\alpha_{k4})^{\frac{11}{2}}} \sum_{i=1}^{N_A} e^{-\frac{\alpha_{k4}}{1+\alpha_{k4}}r_i^2} (x_i + iy_i)^3 z_i \\ c_{n44} &= \sum_{k=1}^{N_k} \frac{3\sqrt{35}\pi\beta_{nk}}{16\sqrt{2}(1+\alpha_{k4})^{\frac{11}{2}}} \sum_{i=1}^{N_A} e^{-\frac{\alpha_{k4}}{1+\alpha_{k4}}r_i^2} (x_i + iy_i)^4 \end{split}$$

$$\begin{split} c_{n50} &= \sum_{k=1}^{N_k} \frac{\sqrt{17} \pi \beta_{nk}}{(16(1+\alpha_{k0})^{\frac{1}{2}})^{\frac{1}{2}}} \sum_{i=1}^{N_k} e^{-\frac{\alpha_{k0}}{1+\alpha_{k0}} r_i^2} z_i (asz_i^4 - 70z_i^2 r_i^2 + 15r_i^4)} \\ c_{n51} &= -\sum_{k=1}^{N_k} \frac{\sqrt{165} \pi \beta_{nk}}{(16\sqrt{2}(1+\alpha_{k0})^{\frac{1}{2}})^{\frac{1}{2}}} \sum_{i=1}^{N_k} e^{-\frac{\alpha_{k0}}{1+\alpha_{k0}} r_i^2} (x_i + iy_i) (21z_i^4 - 14z_i^2 r_i^2 + r_i^4)} \\ c_{n52} &= \sum_{k=1}^{N_k} \frac{\sqrt{155} \pi \beta_{nk}}{8\sqrt{2}(1+\alpha_{k0})^{\frac{1}{2}}} \sum_{i=1}^{N_k} e^{-\frac{\alpha_{k0}}{1+\alpha_{k0}} r_i^2} (x_i + iy_i)^2 z_i (3z_i^2 - r_i^2)} \\ c_{n53} &= -\sum_{k=1}^{N_k} \frac{\sqrt{13} \pi \beta_{nk}}{32(1+\alpha_{k0})^{\frac{1}{2}}} \sum_{i=1}^{N_k} e^{-\frac{\alpha_{k0}}{1+\alpha_{k0}} r_i^2} (x_i + iy_i)^3 (9z_i^2 - r_i^2)} \\ c_{n54} &= \sum_{k=1}^{N_k} \frac{3\sqrt{37} \pi \beta_{nk}}{16\sqrt{2}(1+\alpha_{k0})^{\frac{1}{2}}} \sum_{i=1}^{N_k} e^{-\frac{\alpha_{k0}}{1+\alpha_{k0}} r_i^2} (x_i + iy_i)^4 z_i} \\ c_{n55} &= -\sum_{k=1}^{N_k} \frac{3\sqrt{77} \pi \beta_{nk}}{32(1+\alpha_{k0})^{\frac{1}{2}}} \sum_{i=1}^{N_k} e^{-\frac{\alpha_{k0}}{1+\alpha_{k0}} r_i^2} (x_i + iy_i)^5 \\ \\ c_{n60} &= \sum_{k=1}^{N_k} \frac{\sqrt{13} \pi \beta_{nk}}{32(1+\alpha_{k0})^{\frac{1}{2}}} \sum_{i=1}^{N_k} e^{-\frac{\alpha_{k0}}{1+\alpha_{k0}} r_i^2} (x_i + iy_i)^5 \\ \\ c_{n61} &= -\sum_{k=1}^{N_k} \frac{\sqrt{13} 65\pi \beta_{nk}}{32(1+\alpha_{k0})^{\frac{1}{2}}} \sum_{i=1}^{N_k} e^{-\frac{\alpha_{k0}}{1+\alpha_{k0}} r_i^2} (x_i + iy_i)^2 (33z_i^4 - 30z_i^2 r_i^2 + 5r_i^4) \\ \\ c_{n62} &= \sum_{k=1}^{N_k} \frac{\sqrt{13} 65\pi \beta_{nk}}{32(1+\alpha_{k0})^{\frac{1}{2}}} \sum_{i=1}^{N_k} e^{-\frac{\alpha_{k0}}{1+\alpha_{k0}} r_i^2} (x_i + iy_i)^2 (33z_i^4 - 30z_i^2 r_i^2 + 5r_i^4) \\ \\ c_{n64} &= \sum_{k=1}^{N_k} \frac{\sqrt{13} 65\pi \beta_{nk}}{32(1+\alpha_{k0})^{\frac{1}{2}}} \sum_{i=1}^{N_k} e^{-\frac{\alpha_{k0}}{1+\alpha_{k0}} r_i^2} (x_i + iy_i)^3 z_i (11z_i^2 - 3r_i^2) \\ \\ c_{n64} &= \sum_{k=1}^{N_k} \frac{3\sqrt{13} 65\pi \beta_{nk}}{32(1+\alpha_{k0})^{\frac{1}{2}}} \sum_{i=1}^{N_k} e^{-\frac{\alpha_{k0}}{1+\alpha_{k0}} r_i^2} (x_i + iy_i)^3 z_i (11z_i^2 - 3r_i^2) \\ \\ c_{n6-6} &= \sum_{k=1}^{N_k} \frac{3\sqrt{1001} \pi \beta_{nk}}{64(1+\alpha_{k0})^{\frac{1}{2}}} \sum_{i=1}^{N_k} e^{-\frac{\alpha_{k0}}{1+\alpha_{k0}} r_i^2} (x_i + iy_i)^2 z_i (429z_i^4 - 69z_i^4 r_i^2 + 315z_i^2 r_i^4 - 5r_i^6) \\ \\ c_{n70} &= \sum_{k=1}^{N_k} \frac{3\sqrt{35} \pi \beta_{nk}}{64(1+\alpha_{k0})^{\frac{1}{2}}} \sum_{i=1}^{N_k} e^{-\frac{\alpha_{k0}}{1+\alpha_{k0}} r_i^2} (x_i + iy_i)^3 (43z_i^4 - 66z_i^2$$

$$\begin{split} c_{n80} &= \sum_{k=1}^{N_k} \frac{\sqrt{17\pi}\beta_{nk}}{256(1+\alpha_{k8})} \sum_{i=1}^{N_k} \sum_{i=1}^{N_k} \frac{\alpha_{n8}}{4\sqrt{2}(1+\alpha_{k8})} \sum_{i=1}^{N_k} \sum_{i=1}^{N_k} \frac{3\sqrt{17\pi}\beta_{nk}}{4\sqrt{2}(1+\alpha_{k8})} \sum_{i=1}^{N_k} \sum_{i=1}^{N_k} \sum_{i=1}^{N_k} \frac{3\sqrt{17\pi}\beta_{nk}}{128(1+\alpha_{k8})} \sum_{i=1}^{N_k} \sum_{i=1}^{N_k} \frac{\alpha_{nk}}{128(1+\alpha_{k8})} \sum_{i=1}^{N_k} \sum_{i=1}^{N_k} \frac{\alpha_{nk}}{128\sqrt{2}} \sum_{i=1}^{N_k} \sum_{i=1}^{N_k} \sum_{i=1}^{N_k} \frac{\alpha_{nk}}{128\sqrt{2}} \sum_{i=1}^{N_k} \sum_{i=1}^{N_k} \sum_{i=1}^{N_k} \frac{\alpha_{nk}}{128\sqrt{2}} \sum_{i=1}^{N_k} \sum_{i=1}^{N_k} \frac{\alpha_{nk}}{128\sqrt{2}$$