

Elasticipy: A Python package for linear elasticity and tensor analysis

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Summary

Elasticipy is a Python library designed to streamline computation and manipulation of elasticity tensors for materials and crystalline materials, taking their specific symmetries into account. It provides tools to manipulate, visualize, and analyze tensors—such as stress, strain and stiffness tensors—simplifying workflows for materials scientists and engineers.

Statement of Need

In continuum mechanics, the deformation of a material is described by the second-order strain tensor (usually denoted ε) whereas the stress is described by the second-order Cauchy's stress tensor (σ). Under the linear elasticity assumption, the relationship between the elastic strain ε and σ , known as the generalized Hooke's law, is given through the fourth-order stiffness tensor C with:

$$\sigma = C : \varepsilon$$

where ":" denotes the tensor product contracted twice, so that:

$$\sigma_{ij} = C_{ijkl} \varepsilon_{kl}$$

In order to simplify the above equations, one usually uses the so-called Voigt notation, which reads:

$$\begin{bmatrix} \sigma_{11} \\ \sigma_{22} \\ \sigma_{33} \\ \sigma_{23} \\ \sigma_{13} \\ \sigma_{12} \end{bmatrix} = \begin{bmatrix} C_{1111} & C_{1122} & C_{1133} & C_{1123} & C_{1113} & C_{1112} \\ & C_{2222} & C_{2233} & C_{2223} & C_{2213} & C_{2212} \\ & & C_{3333} & C_{3323} & C_{3313} & C_{3312} \\ & & & C_{2323} & C_{2313} & C_{2312} \\ & & & & C_{1313} & C_{1312} \\ & & & & & C_{1212} \end{bmatrix} \begin{bmatrix} \varepsilon_{11} \\ \varepsilon_{22} \\ \varepsilon_{33} \\ 2\varepsilon_{23} \\ 2\varepsilon_{13} \\ 2\varepsilon_{12} \end{bmatrix}$$

The values of C depend on the material, whereas its pattern (set of zero-components, or linear relationships between them) depends on the material's symmetry (Nye, 1985), as outlined in Figure 1.

Pymatgen (Ong et al., 2013) provides some built-in functions to work on strain, stress and elasticity but lacks some functionalities about the tensor analysis. Conversely, Elate (Gaillac et al., 2016) is a project dedicated to analysis of stiffness and compliance tensors (e.g. plotting directional engineering constants, such as Young modulus). It is implemented in the Materials Project (Jain et al., 2013). AnisoVis (Healy et al., 2020) is similar to Elate, but works on MATLAB®.

Form of the (s_{ij}) and (c_{ij}) matrices

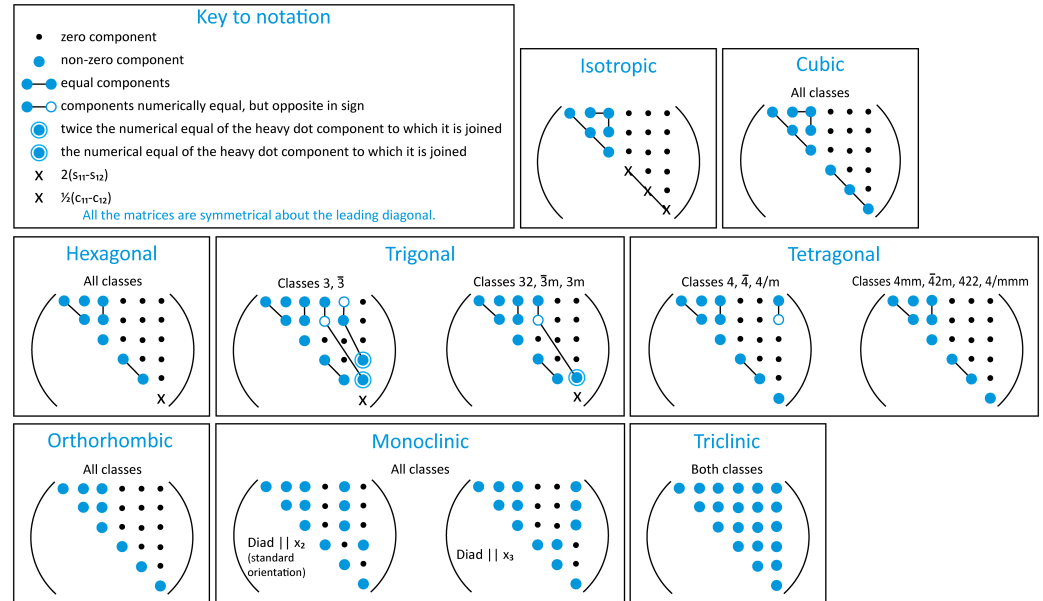


Figure 1: Patterns of stiffness and compliance tensors of crystals, depending on their symmetry (Nye, 1985). With courtesy of Pr. Pamela Burnley.

Therefore, the purpose of Elasticipy is to combine the functionalities of Pymatgen and Elate into a consistent Python project dedicated to continuum mechanics. Its aim is to propose an easy-to-use and efficient tool with the following features:

- intuitive Python-based APIs for defining and manipulating second- and fourth-order tensors, such as strain, stress and stiffness;
- support for standard crystal symmetry groups (Nye, 1985) to facilitate the definition of stiffness/compliance components;
- visualization tools for understanding directional elastic behavior (Young modulus, shear modulus and Poisson ratio);
- a collection of built-in methods to easily and efficiently perform fundamental operations on tensors (rotations (Depriester, 2019), products, invariants, statistical analysis etc.);
- averaging techniques, such as Voigt–Reuss–Hill (Hill, 1952), for textured and non-textured polycrystalline multiphased aggregates.

In order to evidence some of these features, Figure 2.a) illustrates the directional Young modulus of copper (Cu) single crystal as a 3D surface, whereas Figure 2.b) shows the same values as a pole figure (Lambert projection). In Figure 2.c), the Young modulus of a polycrystalline Cu displaying a perfect [001] fiber texture has been estimated with different averaging methods (namely Voigt, Reuss and Hill (Hill, 1952)), then plotted as orthogonal sections.

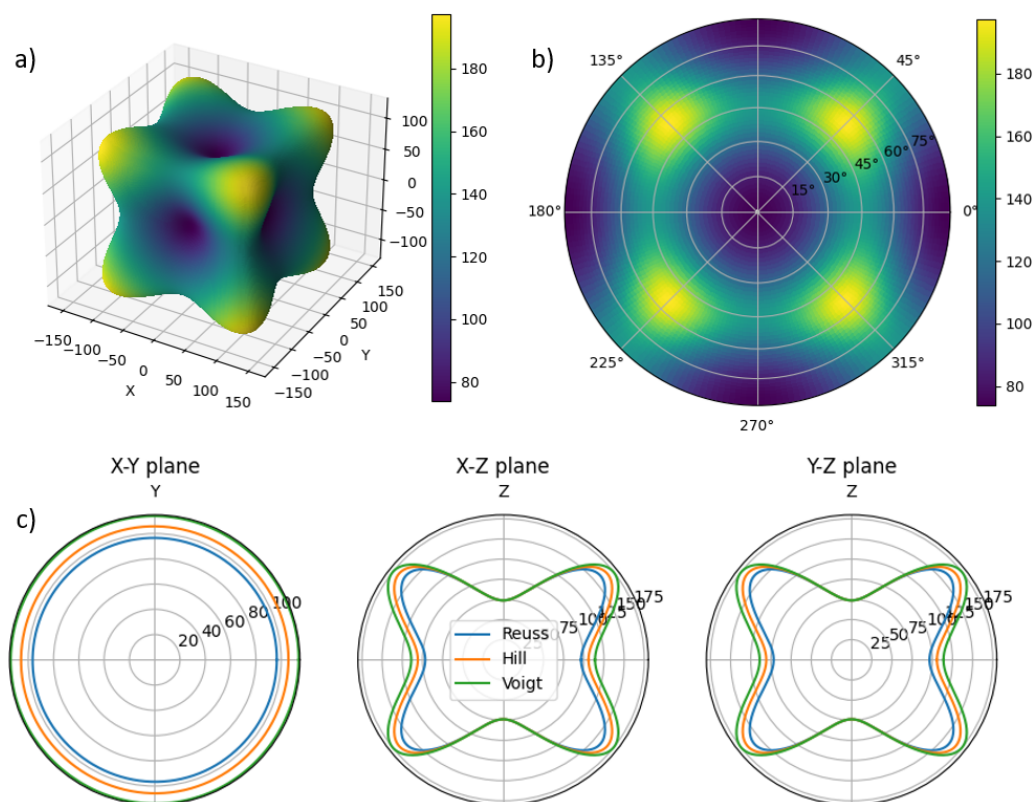


Figure 2: Young modulus (GPa) of Cu single crystal as a 3D surface (a) or a pole figure (b); Young modulus of Cu polycrystal with [001] fiber texture, plotted in three orthogonal sections, depending on the averaging method.

47 Elasticity also introduces the concept of *tensor arrays*, in a similar way as in MTEX (Mainprice
48 et al., 2011), allowing to process several tensors at once with simple and highly efficient
49 commands. In order to highlight the performances of Elasticity, Figure 3 shows the wall-time
50 required to perform two basic operations on tensors (namely, apply the generalized Hooke's law
51 and compute the von Mises equivalent stress) as functions of the number of considered tensors.
52 This demonstrates that, when processing large datasets of tensors ($n > 10^3$), basic tensor
53 operations are 1 to 2 orders of magnitude faster in Elasticity compared to Pymatgen. These
54 performance gains are achieved by leveraging NumPy's array broadcasting capabilities (Harris
55 et al., 2020). However, as tensor algebra is not the primary focus of Pymatgen, Elasticity
56 is designed to complement rather than replace it. Elasticity supports seamless conversion
57 between its own data structures and those of Pymatgen, allowing users to integrate both
58 tools and benefit from Pymatgen's extensive features beyond tensor analysis. Elasticity is also
59 compatible with Orix (Johnstone et al., 2020), a Python library for analysing orientations and
60 crystal symmetry.

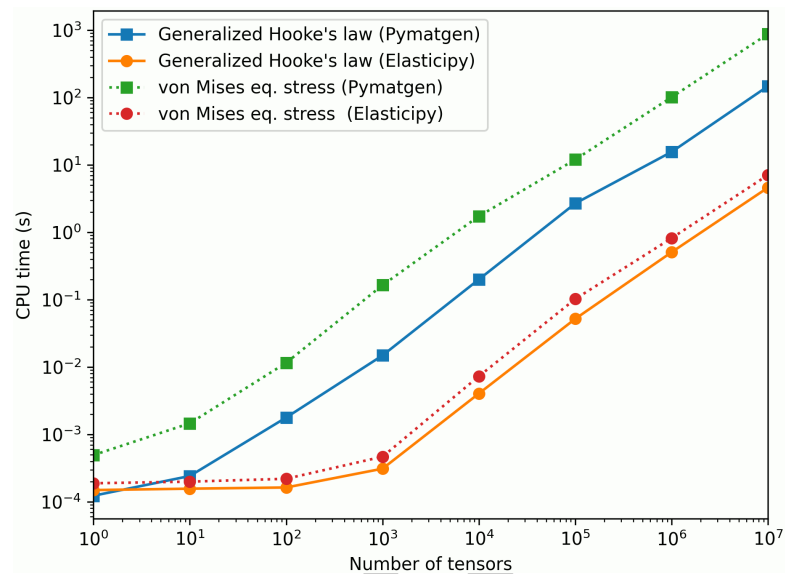


Figure 3: Performance comparison between Elasticity and pymatgen.

Possible extensions

It is worth mentioning that Elasticity provides a full framework for working on tensors, allowing to extend the analyses to other averaging methods (e.g. self-consistent models), possibly beyond linear elasticity problems (e.g. plasticity) with ease. It already implements thermal expansion.

Usage

Plot directional engineering constants

Figure 2.a) and b) were rendered with the following syntax:

```
from Elasticity.tensors.elasticity import StiffnessTensor

C = StiffnessTensor.cubic(C11=186, C12=134, C44=77)
E = C.Young_modulus
E.plot3D(n_phi=500, n_theta=500)
E.plot_as_pole_figure()
```

Create an array of rotated stiffness tensors and compute average

When considering a finite set of orientations, an array of stiffness tensors can be built to account for the rotations:

```
from scipy.spatial.transform import Rotation
import numpy as np

n = 10000
phi1 = np.random.random(n)*2*np.pi # Random sampling from 0 to 2pi
Euler_angles = np.array([phi1, np.zeros(n), np.zeros(n)]).T # Fibre texture
rotations = Rotation.from_euler('ZXZ', Euler_angles) # Bunge-Euler angles
C_rotated = C * rotations # n-length tensor array
```

Then, the Voigt–Reuss–Hill average (Hill, 1952) can be computed from the tensor array:

```
C_VRH = C_rotated.Hill_average()
```

73 Finally, the corresponding Young moduli can be plotted in orthogonal sections, as shown in
74 [Figure 2.c](#)), with:

```
C_VRH.Young_modulus.plot_xyz_sections()
```

75 Arrays of stress/strain tensor

76 Efforts have been made to provide out-of-the-box simple syntaxes for common operations. For
77 example, the following will create a tensor array corresponding to evenly-spaced strain along
78 $[1, 0, 0]$ axis:

```
from Elasticipy.tensors.stress_strain import StrainTensor
```

```
m = 1000 # length of tensor array
mag = np.linspace(0, 0.1, m) # Strain magnitude
strain = StrainTensor.tensile([1, 0, 0], mag)
```

79 Given the stiffness tensor C (see above), one can compute the corresponding stress array with:

```
stress = C * strain
```

80 Finally, `stress.von_Mises()` returns an array of length n and data type `float64`, providing
81 all the corresponding von Mises equivalent stresses.

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