

StrainTool - Improving the Mapping of Tectonic Strain in Eurasia

D.Anastasiou^{1,2}, X. Papanikolaou^{1,2}, V. Kapetanidis^{1,3}, V. Tsironi¹, A. Ganas¹,
D. Paradissis²

¹Institute of Geodynamics - National Observatory of Athens

²Dionysos Satellite Observatory - Higher Geodesy Laboratory,
National Technical University of Athens

³Department of Geology, National and Kapodistrian University of Athens

 <https://dsolab.github.io/StrainTool>

 aganastasiou@gmail.com



Presentation Structure

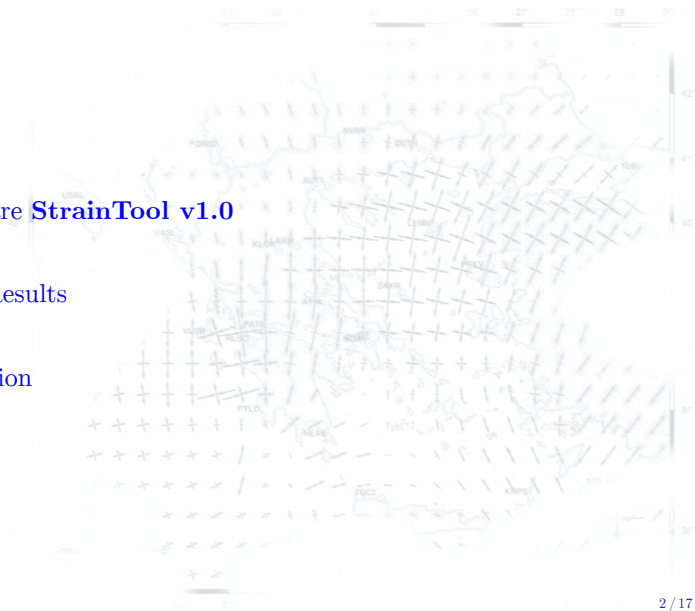
Introduction

Open Source Software **StrainTool v1.0**

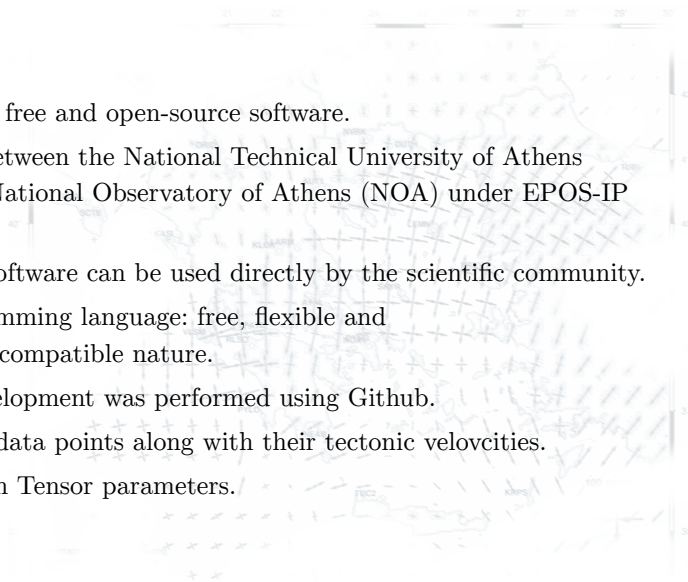
Data analysis and Results

Validation - Discussion

Conclusions



Introduction

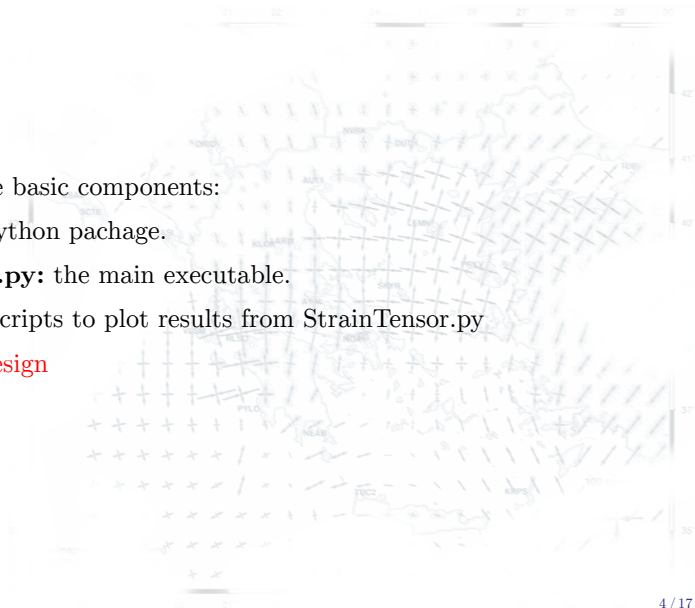
- 
- StrainTool is a free and open-source software.
 - Cooperation between the National Technical University of Athens (NTUA) and National Observatory of Athens (NOA) under EPOS-IP project.
 - User-friendly software can be used directly by the scientific community.
 - Python programming language: free, flexible and cross-platform-compatible nature.
 - Software's development was performed using Github.
 - Input a list of data points along with their tectonic velocities.
 - Estimate Strain Tensor parameters.

Open Source Software **StrainTool v1.0**

StrainTool has three basic components:

- **pystrain:** A python package.
- **StrainTensor.py:** the main executable.
- A list of shell scripts to plot results from StrainTensor.py

TODO: structure design



Python Package **pystrain**

pystrain the core part of the project.

Python functions and classes, enable computation of strain tensor.

The package includes:

- **iotools**: input/output classes to parse ASCII files.
- **geodesy**: functions for basic geodetic calculations.
- **grid.py**: a simple grid generator
- **strain.py**: main class and necessary functions for estimation of strain tensor parameters

Estimate strain tensor parameters

Strain tensor parameters are estimated (or calculated) by solving for the system:

$$\begin{bmatrix} V_{x,S_1} \\ V_{y,S_1} \\ \dots \\ V_{x,S_n} \\ V_{y,S_n} \end{bmatrix} = \begin{bmatrix} 1 & 0 & \Delta_{y_1} & \Delta_{x_1} & \Delta_{y_1} & 0 \\ 0 & 1 & -\Delta_{x_1} & 0 & \Delta_{x_1} & \Delta_{y_1} \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 1 & 0 & \Delta_{y_n} & \Delta_{x_n} & \Delta_{y_n} & 0 \\ 0 & 1 & -\Delta_{x_n} & 0 & \Delta_{x_n} & \Delta_{y_n} \end{bmatrix} \begin{bmatrix} U_x \\ U_y \\ \omega \\ \tau_x \\ \tau_{xy} \\ \tau_y \end{bmatrix}$$

$\Delta_{x_i}, \Delta_{y_i}$ are the displacement components between station i and the point.

A minimum of three stations is required to compute the parameters.

Estimate strain tensor parameters

Assuming that there is a variance information for the station velocities (and a Gaussian distribution), we can add the covariance matrix C of the velocity data in the system. In the simplest case, C is a diagonal matrix, with the velocity component standard deviations as its elements.

$$C = \sigma_0^2 \begin{bmatrix} \left(\frac{1}{\sigma_{V_{x_1} S_1}}\right)^2 & 0 & 0 & \dots & 0 \\ 0 & \left(\frac{1}{\sigma_{V_{y_1} S_1}}\right)^2 & 0 & \dots & 0 \\ 0 & 0 & \left(\frac{1}{\sigma_{V_{x_2} S_2}}\right)^2 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & \left(\frac{1}{\sigma_{V_{y_i} S_i}}\right)^2 \end{bmatrix}$$

Shen Algorithm

Shen et al. 2015, propose a more elaborate approach.

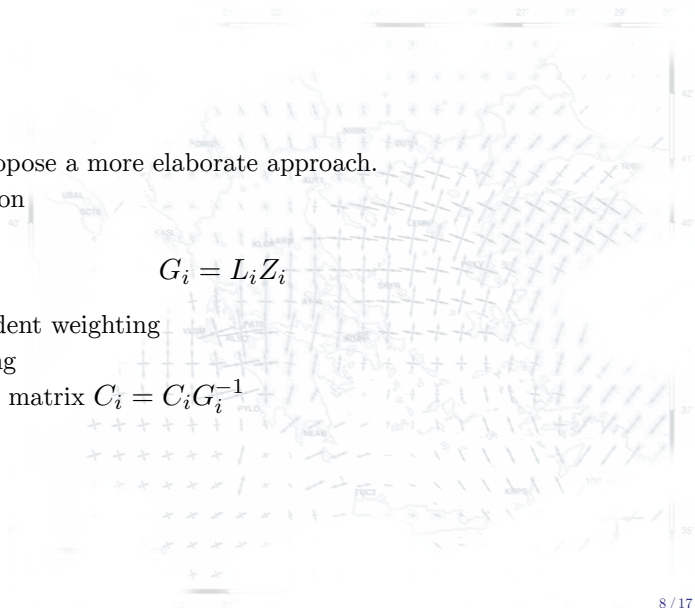
The weighing function

$$G_i = L_i Z_i$$

L_i : distance-dependent weighting

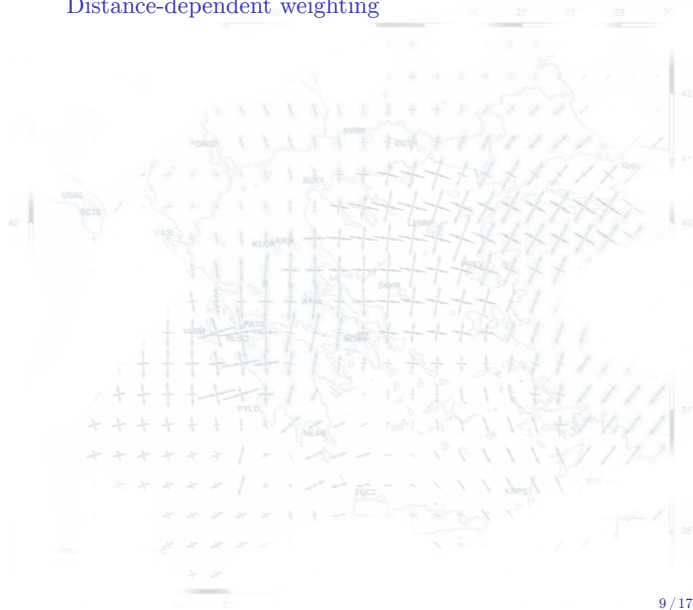
Z_i : spatial weighting

The final covariance matrix $C_i = C_i G_i^{-1}$



Shen Algorithm

Distance-dependent weighting



Shen Algorithm

Optimal smoothing parameter D

Smoothing coefficient D needs to actually compute the distance-dependent weights L_i .

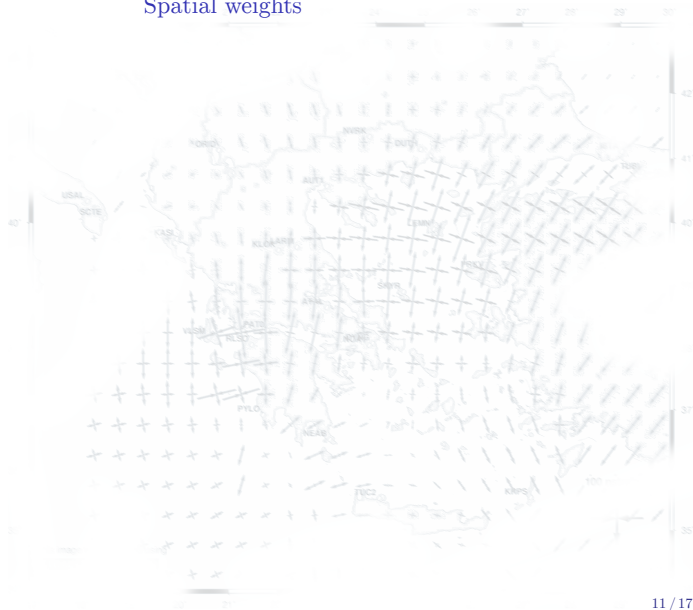
- Either pass in the parameter value as a command line option, or
- search for an optimal D value, given the range Dmin, Dmax, Dstep, and the limit value W

Searching for an optimal D value

1. If a Gaussian approach is selected, then $L_{\max} = 2.15D$ (in km), while for the quadratic approach $L_{\max} = 10D$ (in km).
2. Compute distance-dependent and spatial weights L_i and Z_i respectively, for every station
3. compute the summary $W = 2 \sum Z_i L_i$.
4. repeat the process for the next D value if absolute value W is smaller than W_t , else current D value is optimal.

Shen Algorithm

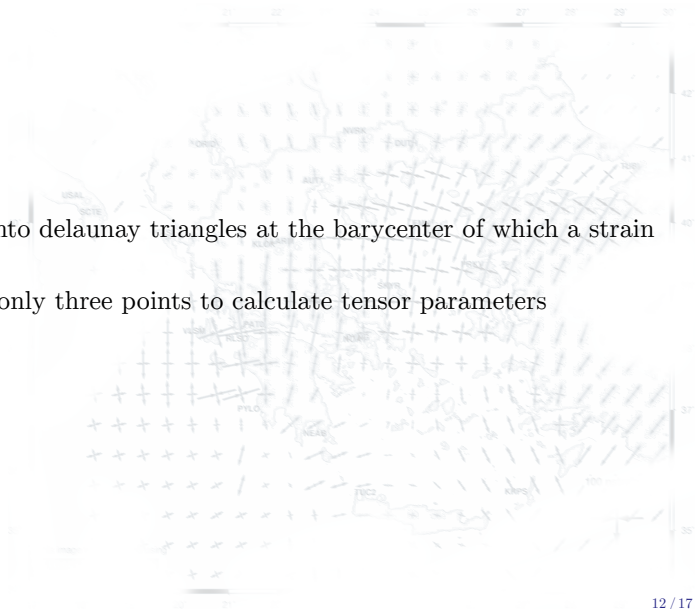
Spatial weights



Veis Algorithm

The region is split into delaunay triangles at the barycenter of which a strain tensor is computed.

This approach uses only three points to calculate tensor parameters



Strain Tensor parameters

$$\tau_{max} = \sqrt{\tau_{xy}^2 + e_{diff}^2}$$

$$e_{max} = e_{mean} + \tau_{max}$$

$$e_{min} = e_{mean} - \tau_{max}$$

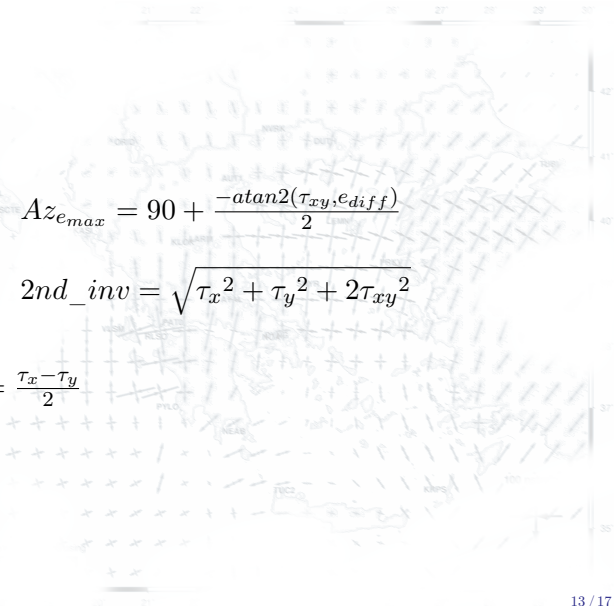
$$dil = \tau_x + \tau_y$$

$$Az_{e_{max}} = 90 + \frac{-atan2(\tau_{xy}, e_{diff})}{2}$$

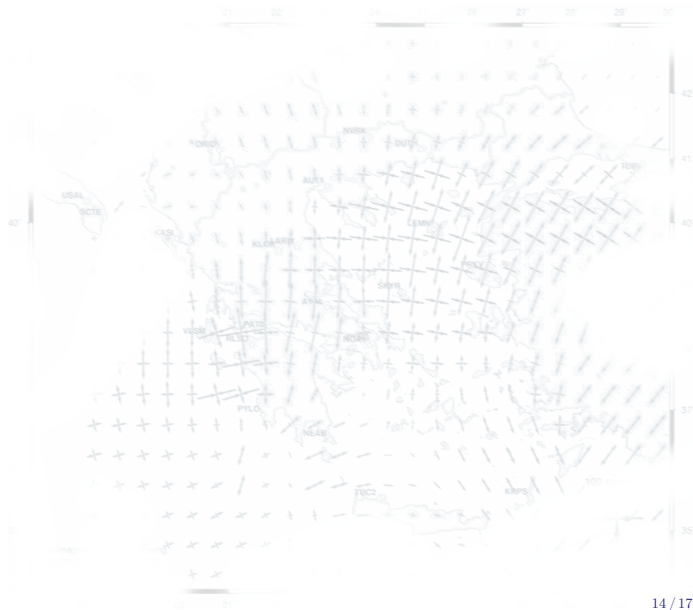
$$2nd_inv = \sqrt{\tau_x^2 + \tau_y^2 + 2\tau_{xy}^2}$$

where,

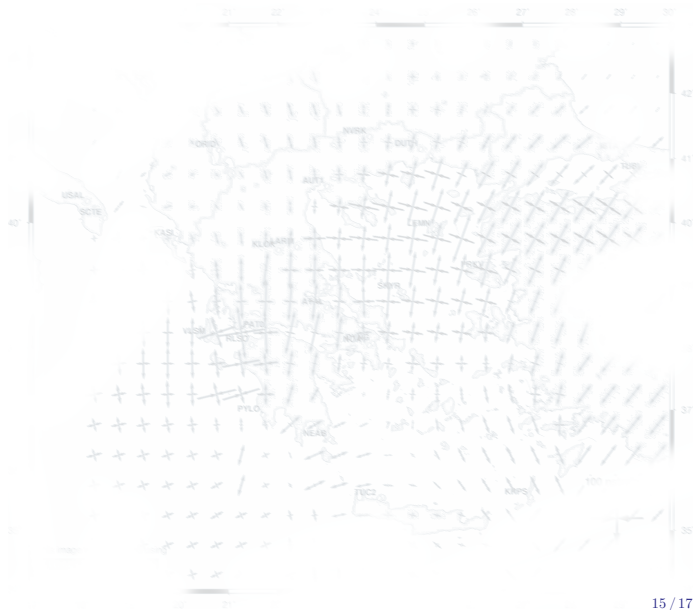
$$e_{mean} = \frac{\tau_x + \tau_y}{2} \quad e_{diff} = \frac{\tau_x - \tau_y}{2}$$



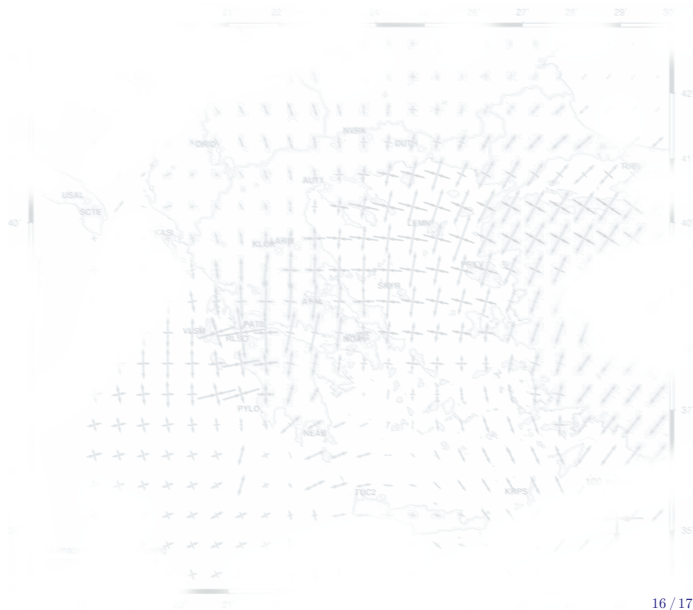
Input Datasets



Validation



Conclusions





Thank you for your attention!

