



MCMTpy: A Python Package for Source Parameters Inversion Based on Cut-And-Paste Algorithm and Markov Chain Monte Carlo

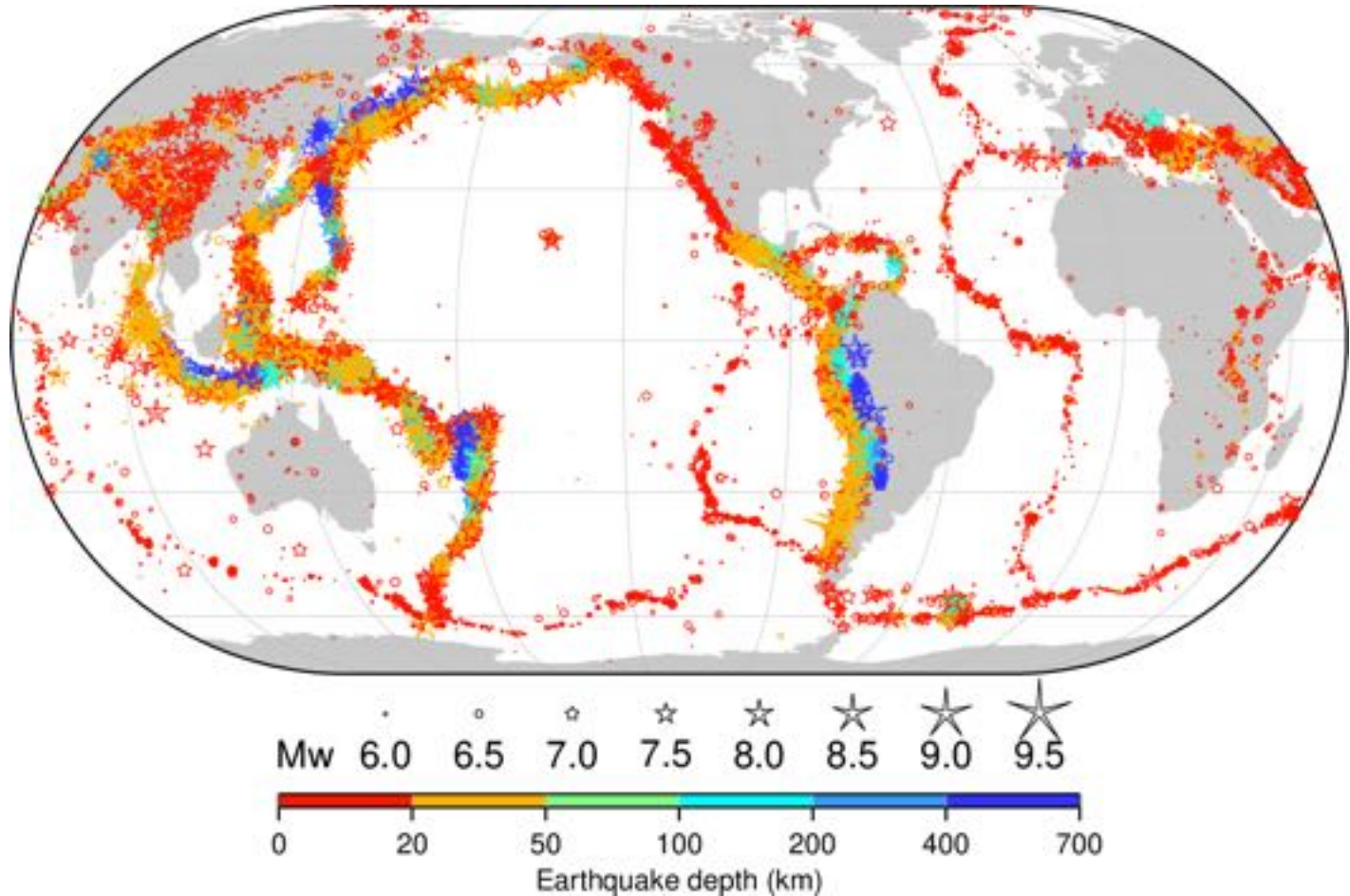
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University of Science and Technology of China

2021.11.01

Global seismicity and plate tectonics

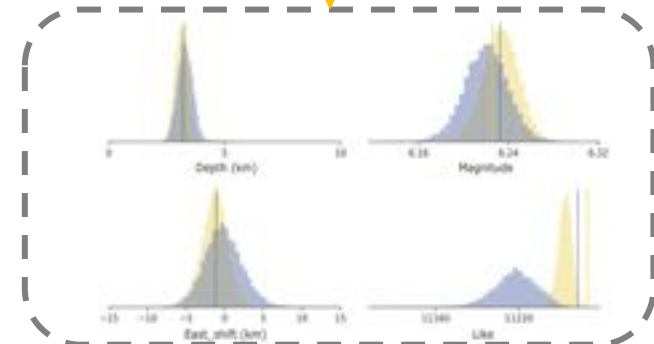
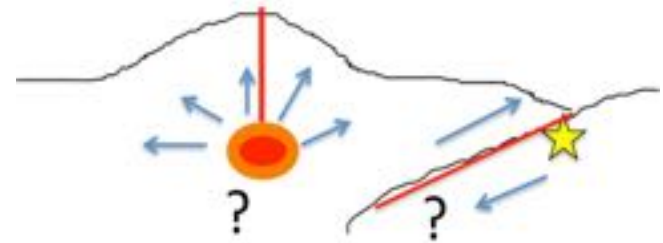


Map showing the earthquakes in version 3.0 of the ISC-GEM catalogue (more than 24,000 earthquakes (Storchak, et al., 2012), <http://www.isc.ac.uk/iscgem/overview.php>).

San Andreas Fault



Inversion Uncertainty Assessment Important



11/1/21

(<https://hvasbath.github.io/beat/index.html>)

CATALOG

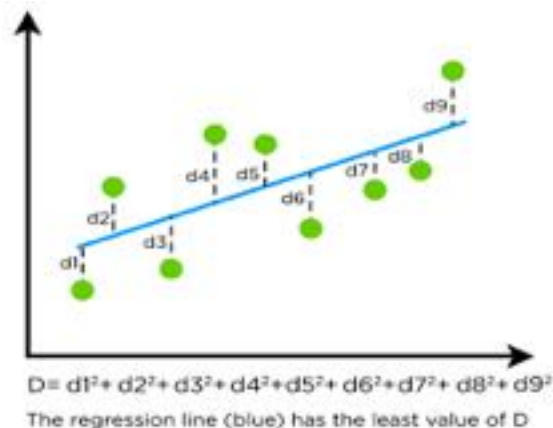
- Introduction
- Method
- Application
- Conclusion

Introduction

Local optimization

- find the minimum of the objective function

$$\min \|\mathbf{G}\mathbf{m} - \mathbf{d}\|_2^2 + \alpha^2 \|\mathbf{m}\|_2^2$$



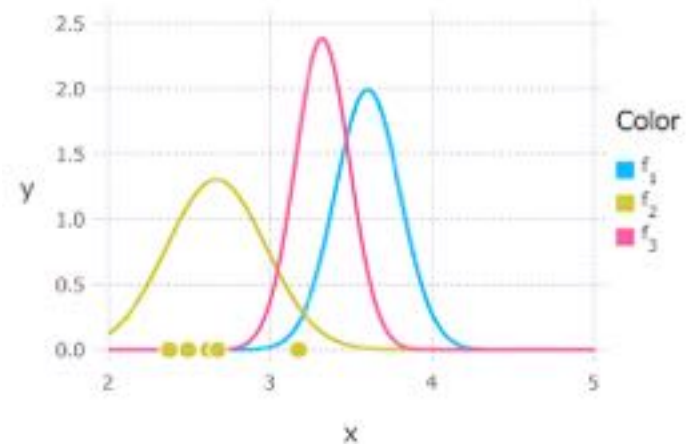
Statistical inverse method (Bayesian)

- find the distribution of model parameters (posterior distribution)

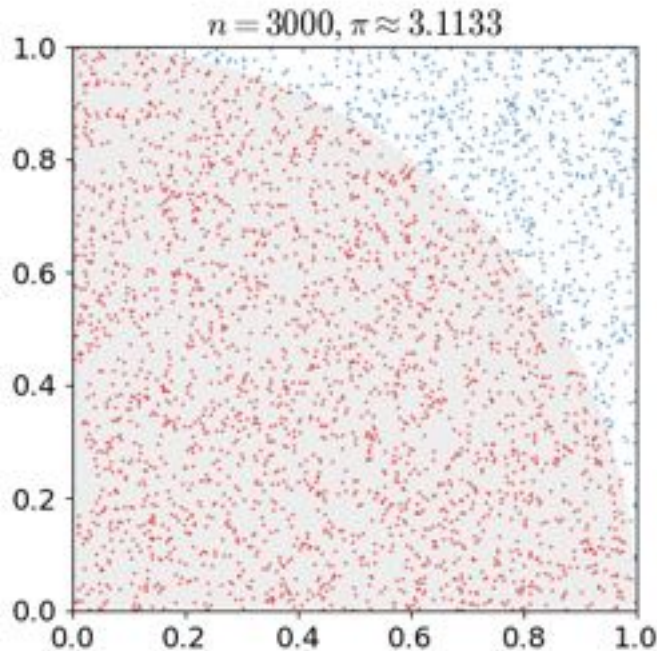
$$\pi_{post}(m|d_{obs}) = k \pi_{prior}(m) \pi_{like}(d_{obs}|m)$$

$$\pi_{post}(m|d_{obs}) \propto \pi_{like}(d_{obs}|m) = \exp\left\{-\frac{1}{2}S(m)\right\}$$

$$S(m) = (d_{obs} - f(m))^T C_k^{-1} (d_{obs} - f(m))$$



Monte Carlo



Monte Carlo method was first proposed by the United States in the 1940s in the Second World War to develop the atomic bomb "Manhattan Project" plan, for secret choice of the gambling city of Monaco Monte Carlo as the code name.



Stanisław Ulam



John von Neumann



Turing

1. Draw a square, then inscribe a quadrant within it
2. Uniformly scatter a given number of points over the square
3. Count the number of points inside the quadrant, i.e. having a distance from the origin of less than 1
4. The ratio of the inside-count and the total-sample-count is an estimate of the ratio of the two areas, $\frac{\pi}{4}$.

Bayesian Framework



Inversion Uncertainty Assessment

Detailed Balance Condition:

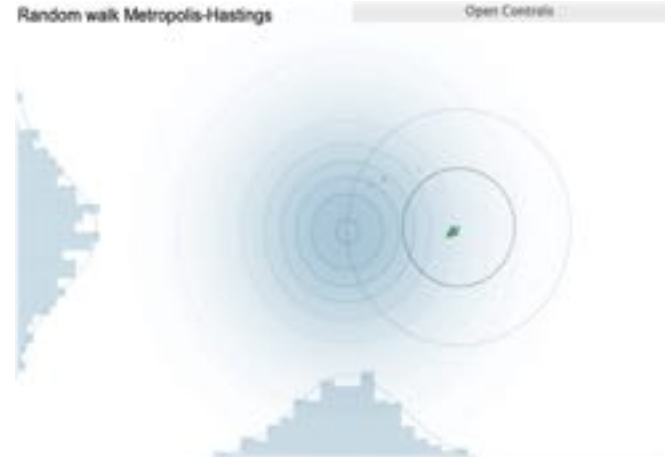
$$\pi(i)P(i, j) = \pi(j)P(j, i), \text{ for all } i, j$$



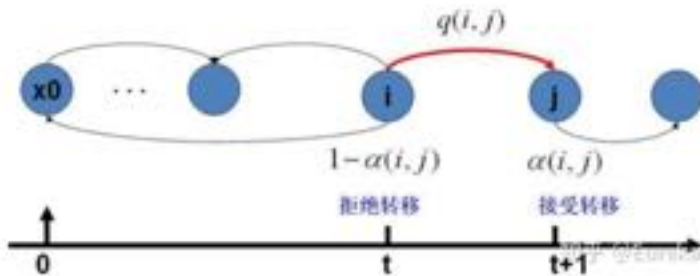
For each transition matrix

$$\pi(i)Q(i, j)\alpha(i, j) = \pi(j)Q(j, i)\alpha(j, i)$$

$\alpha(i, j)$ is acceptance rate



<https://chi-feng.github.io/mcmcdemo/app.html?algorithm=RandomWalkMH&target=banana>



Method



= CAP + MCMC

$$S(m) = \begin{cases} S_{time}(m), & k \leq N_k \\ S_{wave}(m), & k > N_k \end{cases}$$

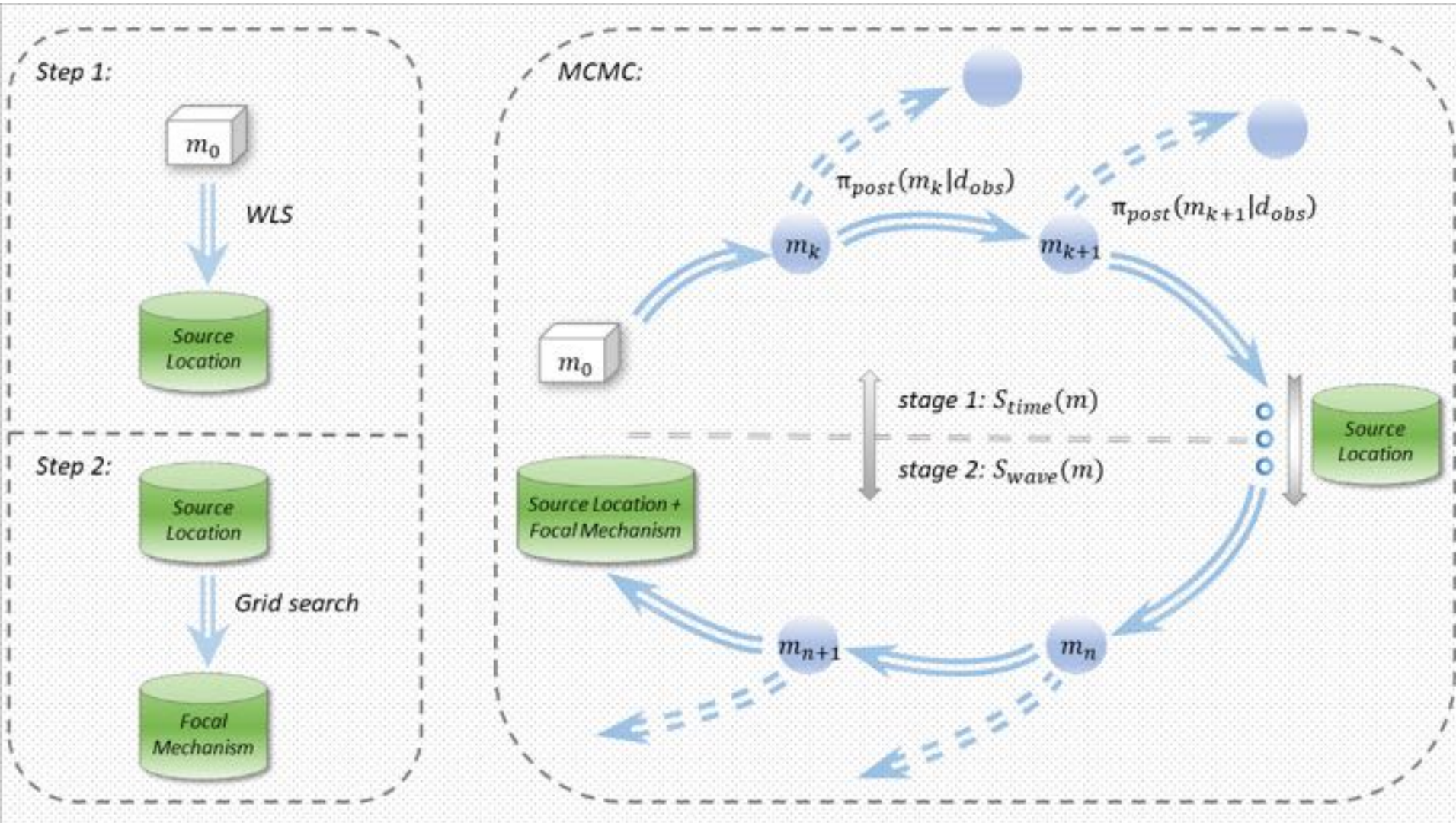
$$S_{time}(m) = \frac{1}{N} \sum_{i=1}^N \frac{[T_i(m) - T_i^{obs} - T_0]^2}{\sigma_i^2},$$

$$S_{waveform} = \frac{1}{N} \sum_{i=1}^N \frac{w_i^{Station}}{\tau_i^2} \left(w_i^p \left(\frac{r_i}{r_0} \right)^{ps_p} (e_i^{p_z} + e_i^{p_r} + e_i^{p_t}) + w_i^s \left(\frac{r_i}{r_0} \right)^{ps_s} (e_i^{s_z} + e_i^{s_r} + e_i^{s_t}) \right. \\ \left. + w_i^{surf} \left(\frac{r_i}{r_0} \right)^{ps_{surf}} (e_i^{surf_z} + e_i^{surf_r} + e_i^{surf_t}) \right),$$

Algorithm 1. MCMTpy algorithm to sample proposal distribution $\pi_{post}(m|d_{obs})$

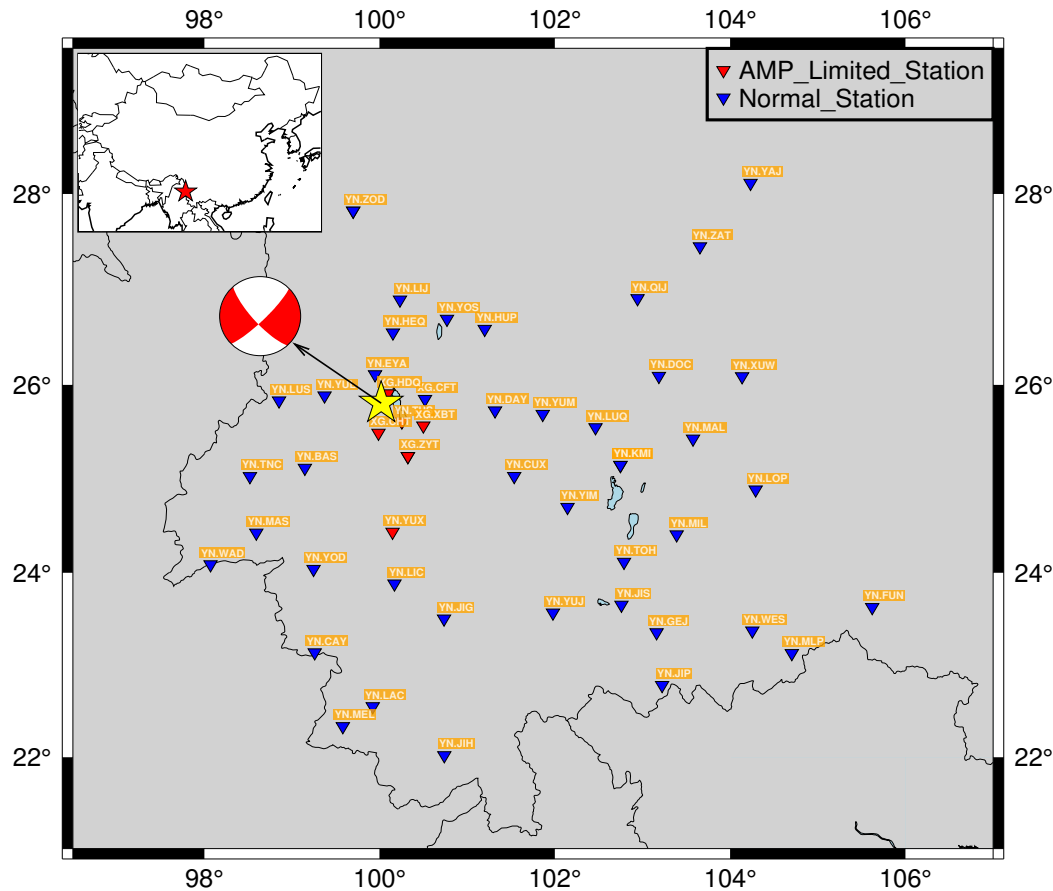
-
1. Choose initial $m_0, S(m_0)$
 2. Compute $\pi_{post}(m_0|d_{obs})$
 3. **for** $k = 0, \dots, N - 1$ **do**
 4. **If** $k < N_k$ **then**
 5. Define $S(m) = S_{time}(m)$
 6. **else**
 7. **If** $k < N_k + N_{mag}$ **then**
 8. Estimate M_0 with formula 15
 9. **end if**
 10. Define $S(m) = S_{wave}(m)$
 11. **end if**
 12. Draw sample y with random walk with formula 18
 13. Compute $\pi_{post}(y|d_{obs})$
 14. Compute $\beta(m_k, y) = \min \left\{ \frac{\pi_{post}(y|d_{obs})}{\pi_{post}(m_k|d_{obs})}, 1 \right\}$
 15. Draw random number $u \sim \mu([0, 1])$
 16. **If** $u < \beta(m_k, y)$ **then**
 17. Accept: set $m_{k+1} = y$
 18. **else**
 19. Reject: set $m_{k+1} = m_k$
 20. **end if**
 21. **end for**
-

Two step workflow **VS** MCMTPy workflow



Application

Application to the 21 May 2021 Ms 6.4 Yangbi Earthquake



According to CENC catalog:

Latitude 25.70°

Longitude 99.88°

Depth 10 km

GMT 13:48:35

Magnitude Ms 6.4

Misfit curve

$$S(m) = \begin{cases} S_{time}(m), & k \leq N_k \\ S_{wave}(m), & k > N_k \end{cases}$$

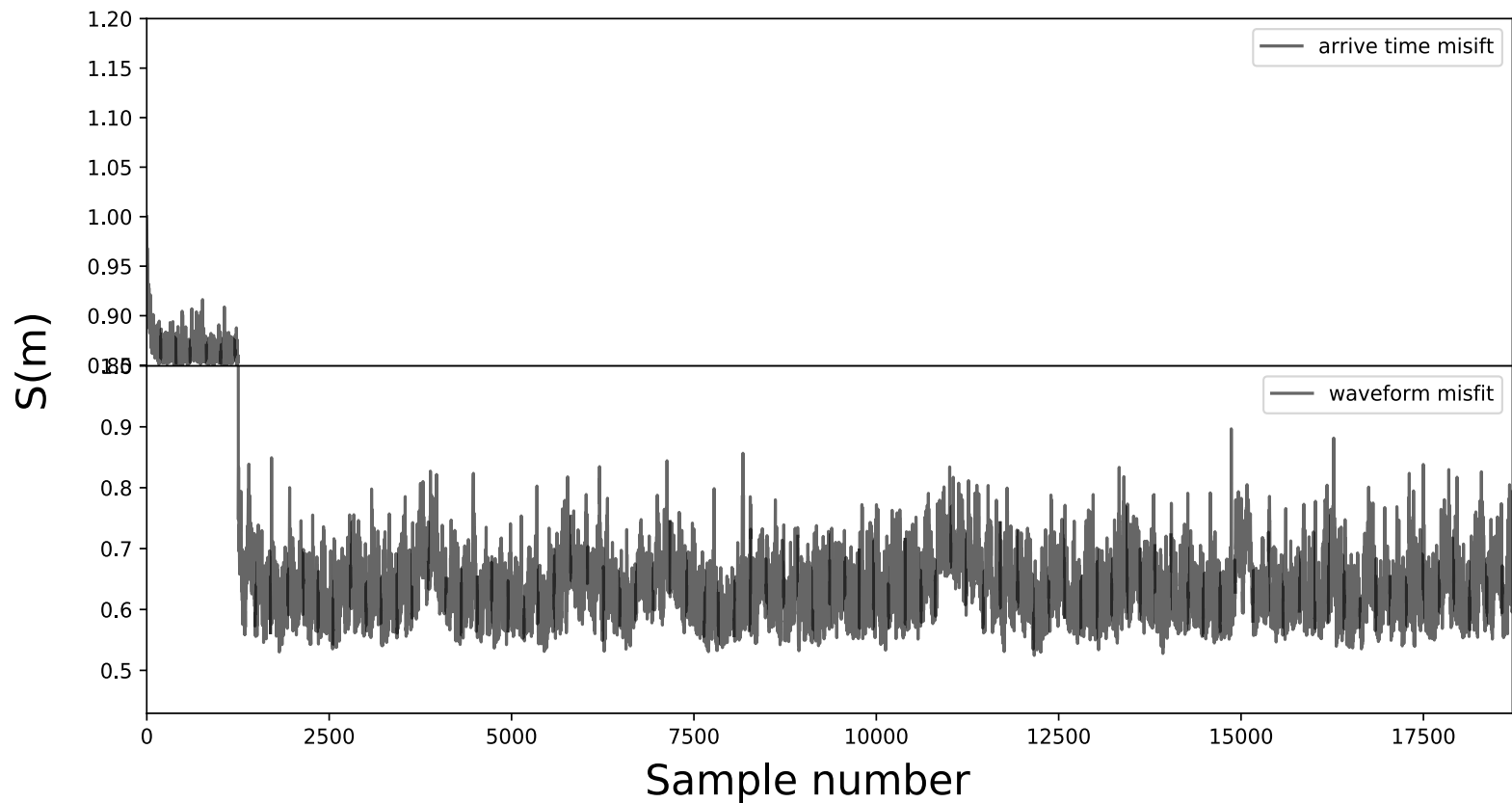







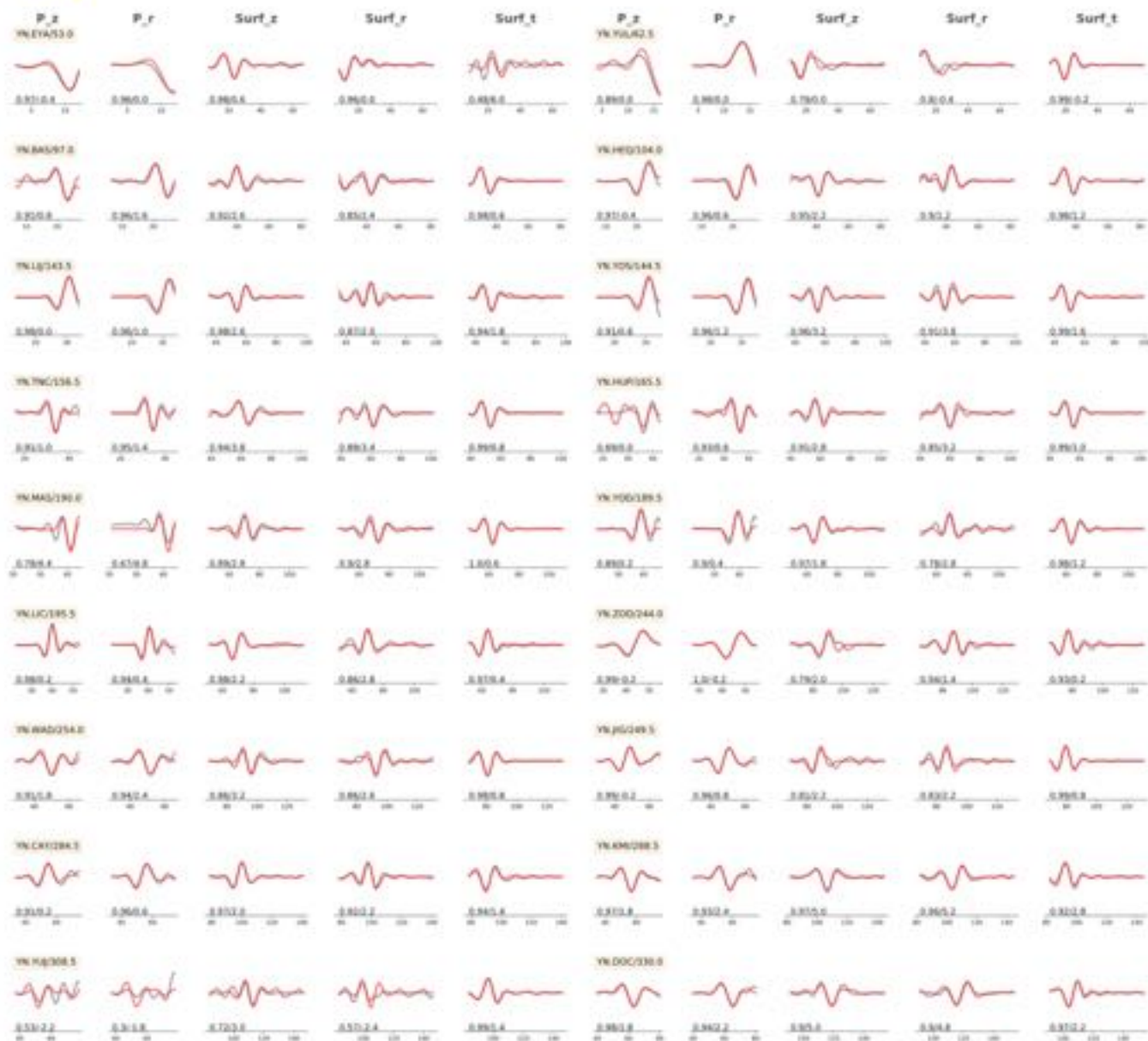
Table 1. Source Parameters for 2021 Yangbi Earthquake of Different Agencies

Method	Mag	Lat/Lon	Depth	Plane 1 and 2 (Strike/Dip/rake)	Percent DC (%)	Beachball
Global CMT	6.1 M_{wc}	25.76/100.01	15.0	315/86/168 46/78/4	62	
USGS W-phase	6.1 M_{ww}	25.74/100.02	17.5	135/82/-165 43/75/-9	93	
P Wave First Motion	NO	25.67/99.87	NO	141/68/-153 40/65/-24	NO	
CAP	6.4 M_s	25.67/99.87	5.0	138/82/-161 45/71/-8	NO	
MCMTpy	6.65 M_w	25.62/99.93	5.2	138/76/-170 46/80/-14	91	

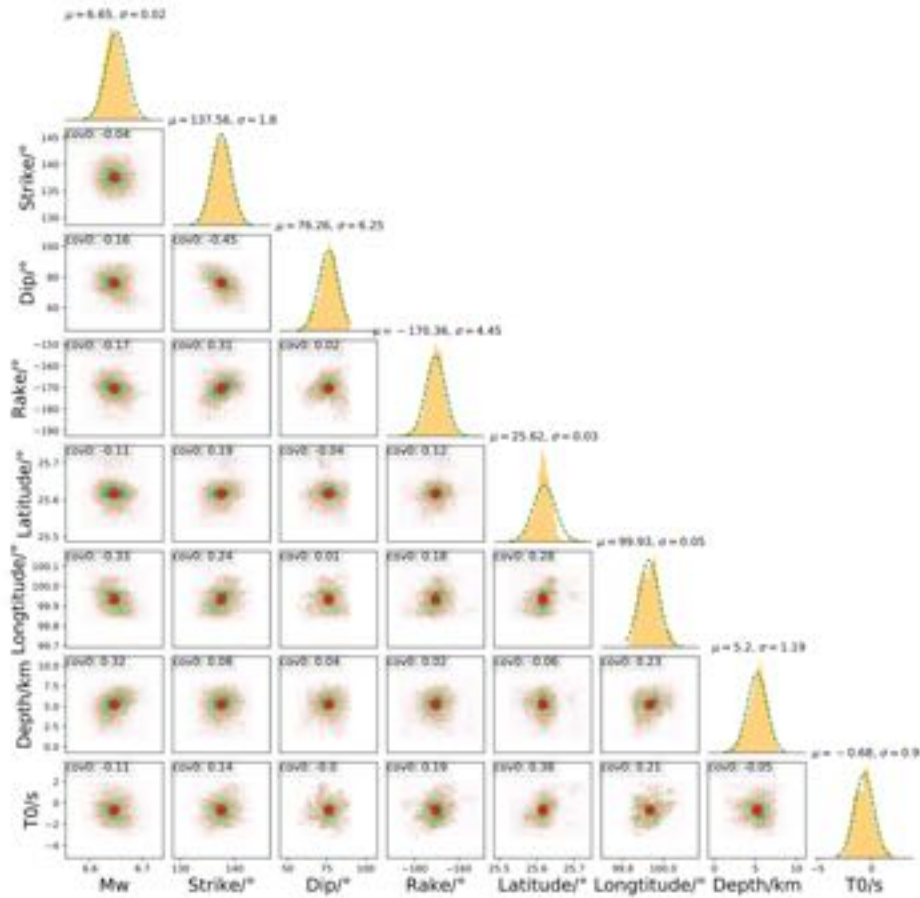
P Wave First Motion and CAP results are from the Institute of Geophysics, China Earthquake Administration, using global network data and local stations, respectively.



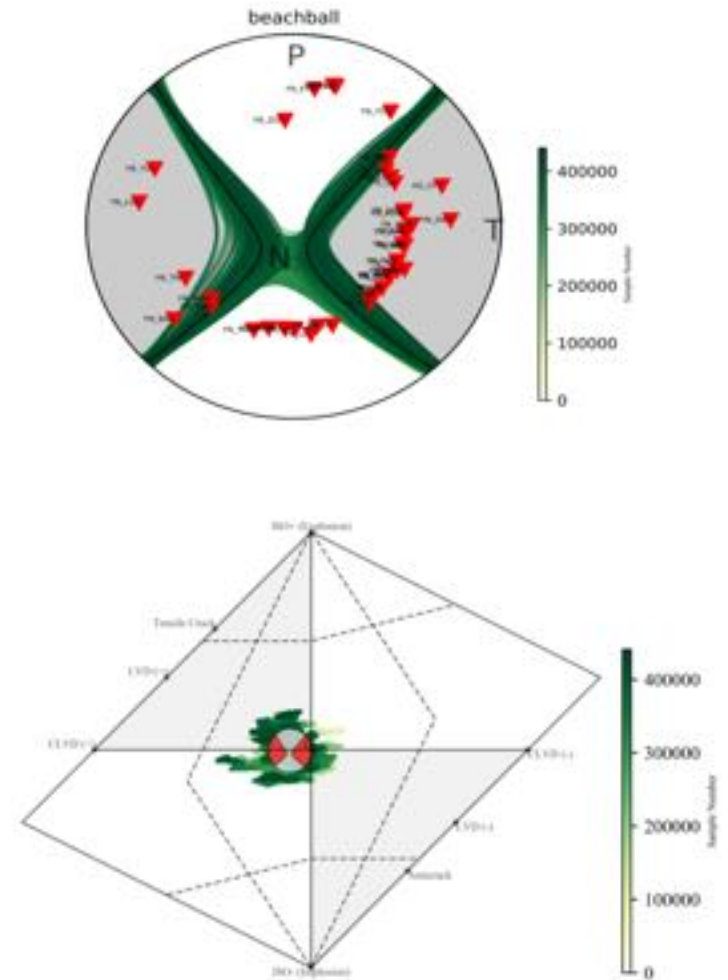
Lat: 25.63 Lon: 99.93 Dep: 5.1 TO: -0.0
Fm: 45.3 79.6 -13.5 Mw: 6.65



Inversion Result

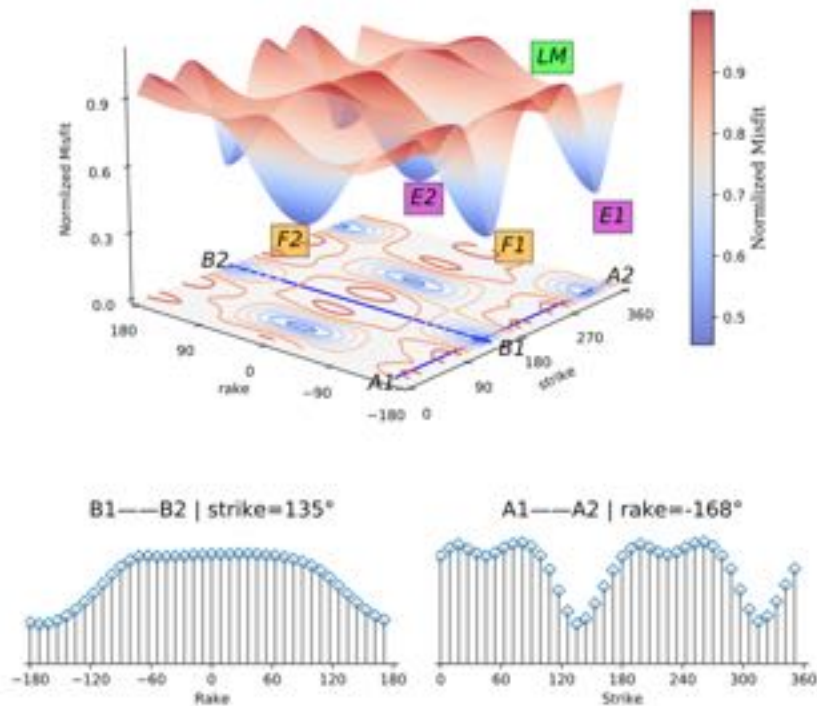


Double Couple Inversion

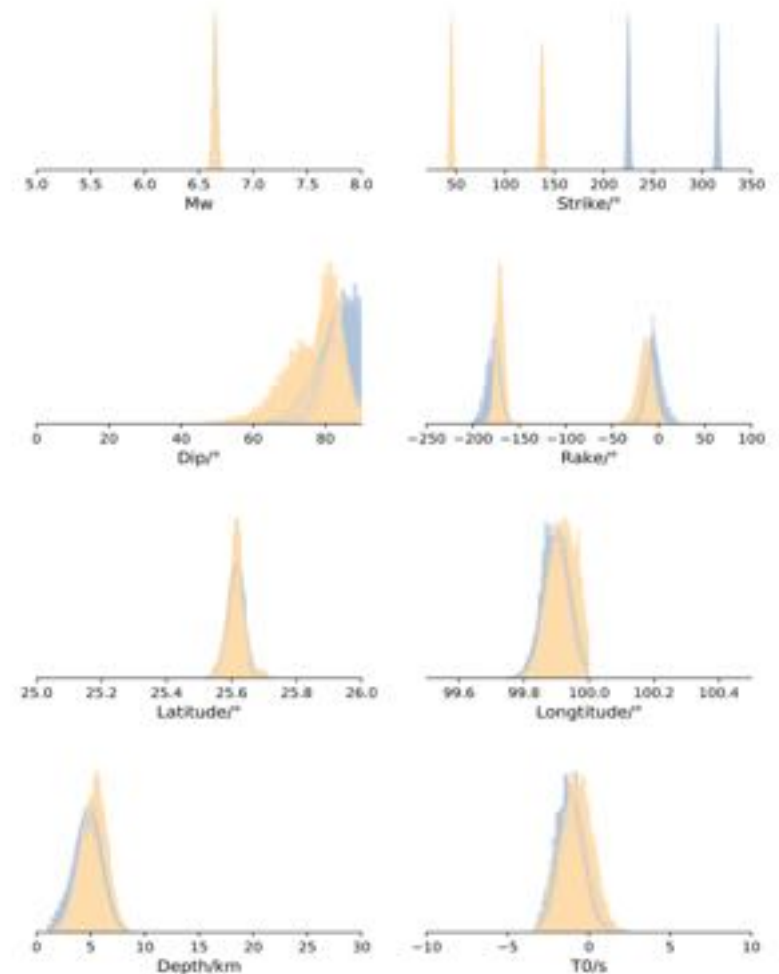


Moment Tensor Inversion

Reduce the Dependence of **Prior Model** and Limitations



two-dimensional normalized misfit space of strike-rake



30 random initial models

- Source parameters inversion under **double-couple** assumption with Markov Chain Monte Carlo method.
- Source parameters inversion under double-couple assumption with **grid search method**.
- Source parameters inversion for full **moment-tensor** solution with Markov Chain Monte Carlo method.
- Jupyter notebooks for source **parameter calculation and conversion**, such as moment tensor decomposition and Hudson plot.



☐ Tutorial_Detailed

Calculate Green Function Database
Synthesize the Test Data
Prepare Data for MCMTPy
Inversion of Source
Result Visualization

☐ Conversion of Source Parameters

The installation

str_dip_rake to mt

The conversion between str_dip_rake and A/N vector

The installation

- jupyter notebook:

```
# import MomentTensor
from MCMTPy import MomentTensor as MTpy

# get __doc__
MTpy.__doc__.strip().split("\n")
```

str_dip_rake to mt

- jupyter notebook:

```
# function: str_dip_rake2MT
MTpy.str_dip_rake2MT.__doc__.strip().split("\n")

# input
strike = 50
dip = 50
rake = 100

A = MTpy.str_dip_rake2MT(strike,dip,rake)
A.mt
```

More Information

MCMTpy

Public

<https://github.com/OUCyf>

MCMTpy is a Python package designed for seismic source study. It provides functionality for focal mechanism inversion and source parameters analysis.

Jupyter Notebook 11 MIT License Updated 15 days ago

README.rst

MCMTpy

About

The package is still undergoing development.

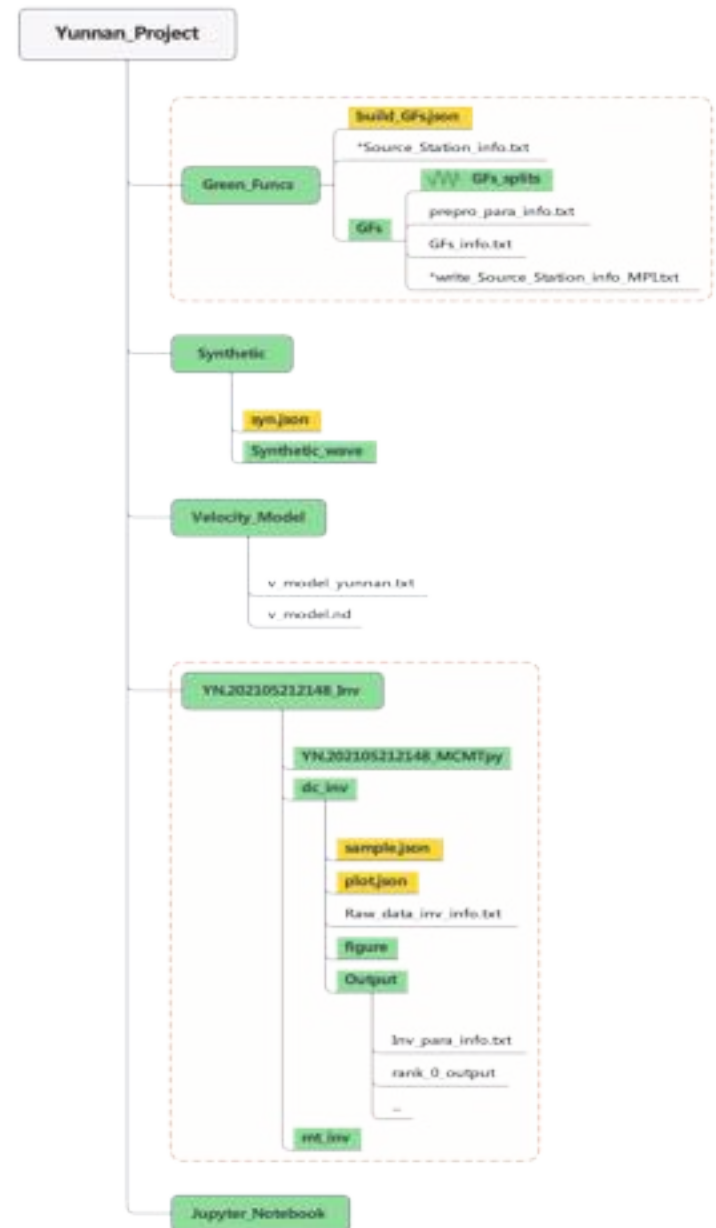
MCMTpy is a Python package designed for seismic source study. It provides functionality for focal mechanism inversion and source parameters analysis.

MCMTpy has mainly provided functions as:

- Focal mechanism inversion of the double couple source and moment tensor using improved Markov chain Monte Carlo method.
- A series of scripts is provided for source parameter calculation and conversion.
- Moment tensor decompose, Huston plot and Beachball with station projected.

and it has the features as:

- The package MCMTpy runs on Unix-like systems including Mac and Linux.
- It has a close integration with Obspy, pyfk and pyasdl.
- It can be done in parallel based on mpi4py.
- It's efficient to do the seismological research based on python.



Conclusion

- ❑ We present the MCMTpy package using the MCMC method to combine source location inversion and focal mechanism inversion on one Markov chain.
- ❑ The new method considers both **phase arrival time** and **waveform information** to effectively reduce the dependence of the prior model and avoid trapping into local minima.
- ❑ This method can mitigate the **cumulative uncertainties** of the source location, provides a way to quantify source parameter's uncertainties by statistical inference.

Thanks !