

2.1 Open Source Interpretability Tools

Using the Example of
FAT Forensics

- Research software and reproducibility.
- Paperware.
- Interpretability, explainability and transparency toolkits.
- FAT Forensics and its design principles.

Research Software

- Lucky to have it.
- Often not maintained.
- Sometimes lengthy Python scripts.
- Heavy and unnecessary dependencies (or simply unavailable).
- Over (or under) engineered.



alexhepburn commented 1 minute ago



Please could you upload `missing-module` so that we can run the test scripts?
Thank you.

Reproducibility

Software underpins much of research, but it tends to lack a good foundation.

- *Vapourware* -- promised but not delivered, i.e., non-existent.
- **Paperware** -- available but difficult to use by people outside of the core research (and/or development) team.
- *Software* -- documented, tested, usable and welcoming.

We need `scikit-learn` for Fairness, Accountability and **Transparency**.

Algorithmic Transparency

Individual explainers:

- LIME
<https://github.com/marcotcr/lime>
- Local surrogates
<https://github.com/axa-rev-research/locality-interpretable-surrogate>
- Anchor
<https://github.com/marcotcr/anchor>
- PyCEbox
<https://github.com/AustinRochford/PyCEbox>

Transparency/Interpretability/Explainability **packages**:

- Microsoft's InterpretML
<https://github.com/interpretml/interpret>
- IBM's AI Explainability 360
<https://github.com/IBM/AIX360>
- Oracle's Skater
<https://github.com/oracle/Skater>
- ELI5
<https://github.com/TeamHG-Memex/eli5>
- Yellowbrick
<https://github.com/DistrictDataLabs/yellowbrick>

FAT Forensics

Algorithmic Fairness, Accountability and Transparency Toolkit



Origin

Creating a piece of software that covers fairness, accountability and transparency.



University of
BRISTOL

THALES

Team:

- Kacper Sokol -- Lead Developer
- Alex Hepburn -- Core Developer
- Peter Flach -- Principal Investigator
- Rafael Poyiadzi -- Developer
- Matthew Clifford -- Developer
- Raul Santos-Rodriguez -- Co-Investigator

Design and Development Principles

- Open sourced under the BSD 3-Clause licence.
- Minimal dependencies.
- Good software engineering practices:
 - unit testing;
 - continuous integration; and
 - consistent code styling and formatting.
- Complete and diverse documentation:
 - API reference;
 - online tutorials;
 - how-to guides; and
 - code examples.

Software	license BSD-3-Clause release v0.1.0 pypi v0.1.0 python 3.5
Docs	homepage read
CI	build passing codecov 100%
Try it	launch binder
Contact	mailing list Google Groups chat on gitter
Cite	cite BibTeX JOSS 10.21105/joss.01904 DOI 10.5281/zenodo.3833199

Scope

Fairness

Data

Do some data points share the same unprotected features but different protected features?

Models

Is there demographic parity between certain sub-groups?

Predictions

Are two data points that differ only in protected features treated differently?

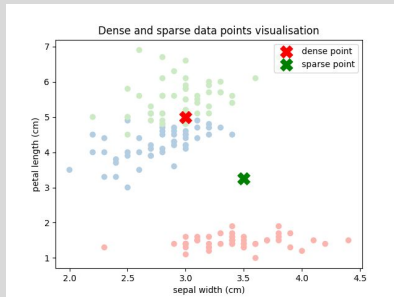
Accountability

Data

Is there a sampling bias in the data according to some sub-groups?

Models

Is there a systematic performance bias in the model?



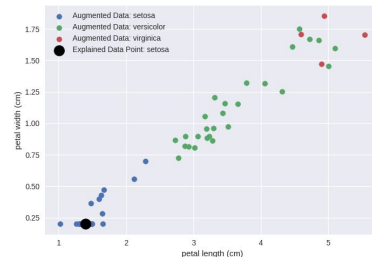
Transparency

Predictions

Why is a decision made?

Models

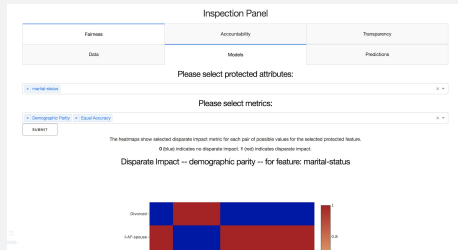
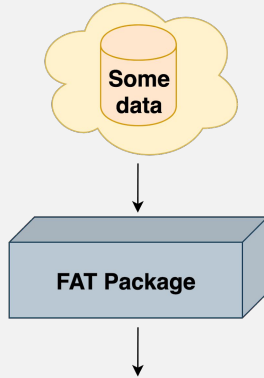
What influence does each feature have on the model?



Use Modes

Deployment Mode

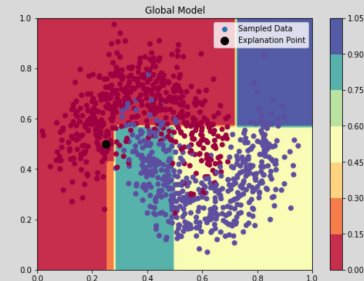
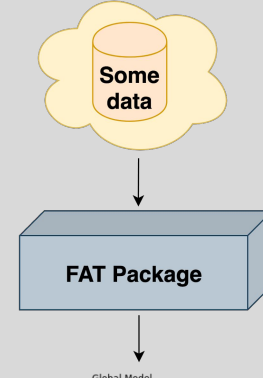
data in -- data out



<https://fatf.herokuapp.com/>

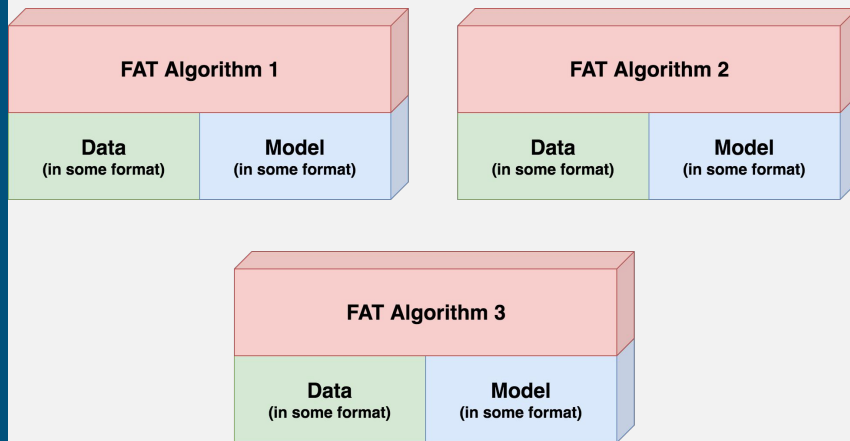
Research Mode

data in -- visualisations out

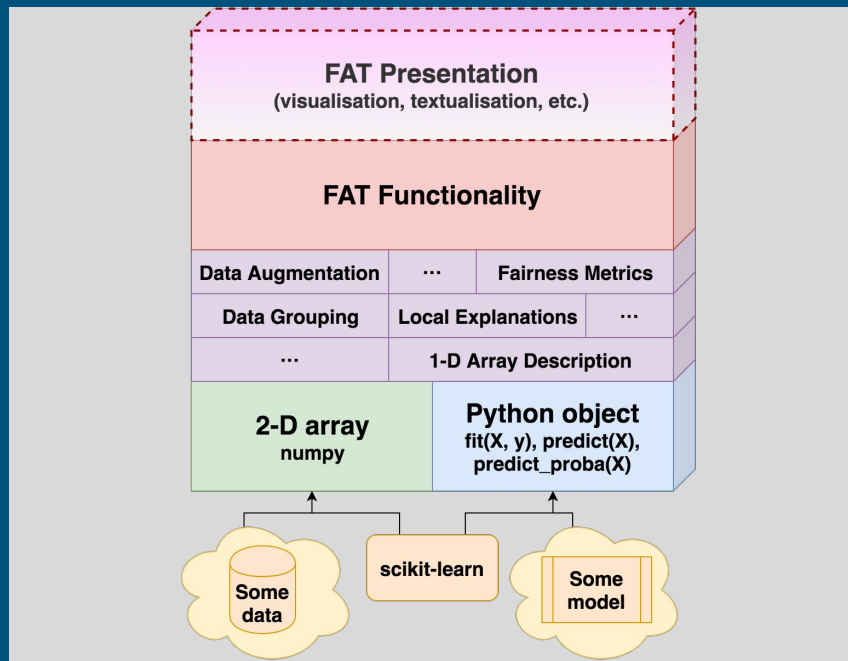


Modularity

Bespoke Code



Modular Design



Implemented Functionality

	Fairness	Accountability	Transparency
Data/ Features	<ul style="list-style-type: none">• Systemic Bias (disparate treatment labelling).• Sample size disparity (e.g., class imbalance).	<ul style="list-style-type: none">• Sampling bias.• Data Density Checker.	<ul style="list-style-type: none">• Data description.
Models	<ul style="list-style-type: none">• Group-based fairness (disparate impact).	<ul style="list-style-type: none">• Systematic performance bias.	<ul style="list-style-type: none">• Partial dependence.• Individual conditional expectation.
Predictions	<ul style="list-style-type: none">• Counterfactual fairness (disparate treatment).		<ul style="list-style-type: none">• Counterfactuals.• Tabular bLIMEy (LIME alternative).

Planned Transparency Features

	Fairness	Accountability	Transparency
Data/ Features			<ul style="list-style-type: none">• Bespoke Interpretable Representations
Models			<ul style="list-style-type: none">• Permutation Importance• Decision Tree Explainer
Predictions			<ul style="list-style-type: none">• Anchors• Image and Text Surrogates• Tree-specific Counterfactuals

<https://github.com/fat-forensics/fat-forensics>

The screenshot shows the GitHub repository page for `fat-forensics/fat-forensics`. The repository is owned by `fat-forensics` and has 4 watchers, 32 stars, and 9 forks. The repository is in the `master` branch, with 8 branches and 3 tags. The repository description is "Modular Python Toolbox for Fairness, Accountability and Transparency Forensics". The repository is licensed under the BSD-3-Clause License. The repository has 39 commits, with the latest commit being `6fa252a` on 19 May. The repository is categorized under `machine-learning`, `fairness`, `accountability`, `transparency`, `interpretability`, `explainability`, `explainable-ai`, and `interpretable-ai`. The repository has a README file and 2 releases. The repository is categorized under `Releases` and `Packages`.

Search or jump to...

Pull requests Issues Marketplace Explore

fat-forensics / fat-forensics

Unwatch 4 Star 32 Fork 9

<> Code Issues Pull requests 7 Actions Security Insights Settings

master 8 branches 3 tags

Go to file Add file Code

So-Cool Updated citation guidelines 6fa252a on 19 May 39 commits

.github	Preparation for open-sourcing	12 months ago
build_tools	0.0.2 documentation update and documentation deployment fix (remove o...	10 months ago
doc	Updated citation guidelines	3 months ago
examples	0.1.0 release and JOSS publication (#32)	4 months ago
fatf	0.1.0 release and JOSS publication (#32)	4 months ago
.coveragerc	Travis yaml, linting and flake8 (close #4)	2 years ago
.editorconfig	Dev environment setup (#20)	2 years ago
.flake8	Package documentation for version 0.0.1 release (#28)	13 months ago
.gitignore	Tabular Surrogates (#29 and #26)	10 months ago
.mypy.ini	0.1.0 release and JOSS publication (#32)	4 months ago
.pylintrc	Dev environment setup (#20)	2 years ago
.style.yapf	Dev environment setup (#20)	2 years ago
.travis.yml	Travis CI skip_cleanup instead of cleanup -- still deploy v1	4 months ago

About

Modular Python Toolbox for Fairness, Accountability and Transparency Forensics

fat-forensics.org

machine-learning fairness accountability transparency interpretability explainability explainable-ai interpretable-ai

Readme


BSD-3-Clause License


Releases 3


FAT-Forensics 0.1.0 Latest on 19 May + 2 releases

Packages

<https://fat-forensics.org/>

 **FAT Forensics** [Home](#) [Documentation](#) [FAT User Guide](#)

ENHANCED BY 



Welcome to FAT Forensics!

FAT Forensics is a Python toolkit for evaluating Fairness, Accountability and Transparency of Artificial Intelligence systems. It is built on top of [SciPy](#) and [NumPy](#), and distributed under the 3-Clause BSD license (new BSD).

In addition to the code documentation, this web page also includes a detailed [User Guide](#) that describes FAT algorithms on a more theoretical level and talks about best practices when using them.

A great way to get yourself familiar with the package and where it comes from is the [Getting Started](#) page.

Source Code

For hosting the source code we use the [FAT-Forensics](#) organisation on GitHub, with the source code for the FAT Forensics packed being held in the [fat-forensics](#) repository.

Communication

We use a range of platforms for communication in the project:


- for issues with the source code or the documentation please open an issue on our [GitHub issue tracker](#);
- the code-related development discussion should happen on our [gitter](#) channel;
- the discussion about the project's future and the direction of the development happens on our [slack](#) channel and [mailing list](#).

Acknowledgement

The project has started as an academic collaboration between the [University of Bristol](#) and Thales. You can find all of our contributors and more information about the support we receive [here](#).

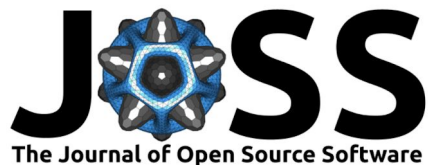
Please remember to [cite us](#) if you use any part of the package or its documentation.

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 **THALES**
More information on our contributors

[Show this page source](#)

<https://joss.theoj.org/papers/10.21105/joss.01904>



FAT Forensics: A Python Toolbox for Implementing and Deploying Fairness, Accountability and Transparency Algorithms in Predictive Systems

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<https://arxiv.org/abs/1909.05167>

**FAT Forensics:
A Python Toolbox for Algorithmic Fairness,
Accountability and Transparency**

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Hands-on Session Preparation

(Alex Hepburn)