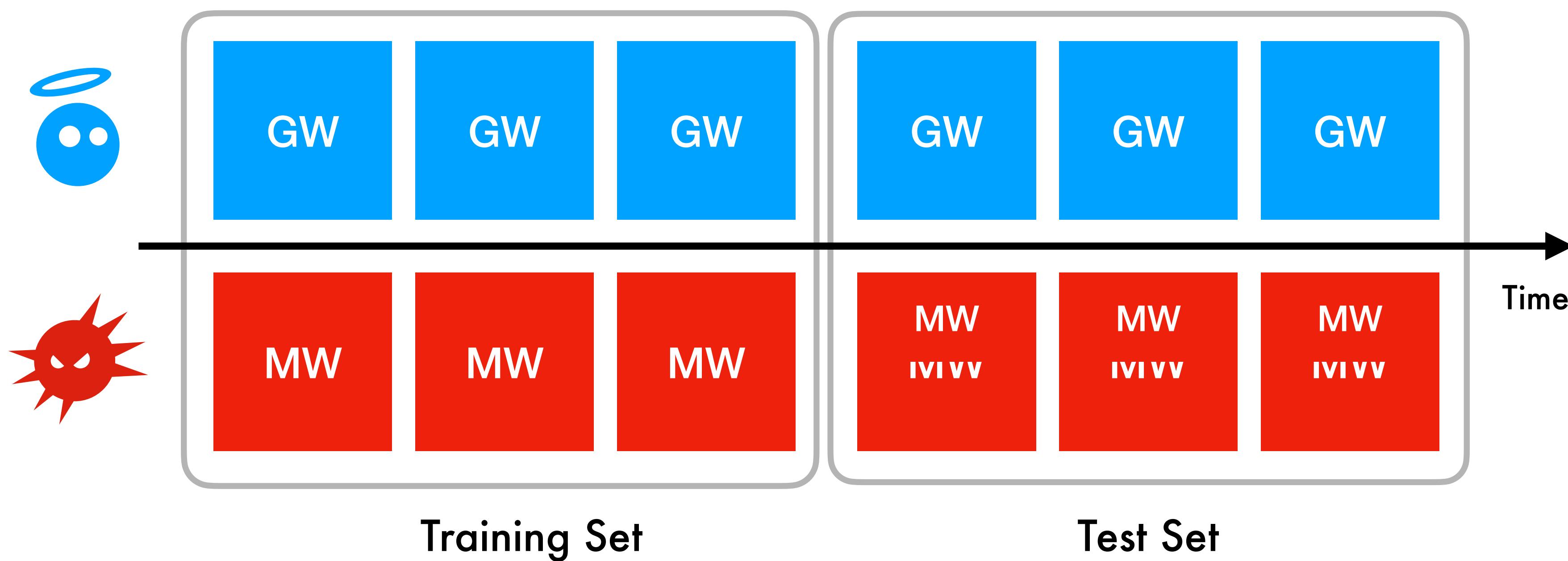
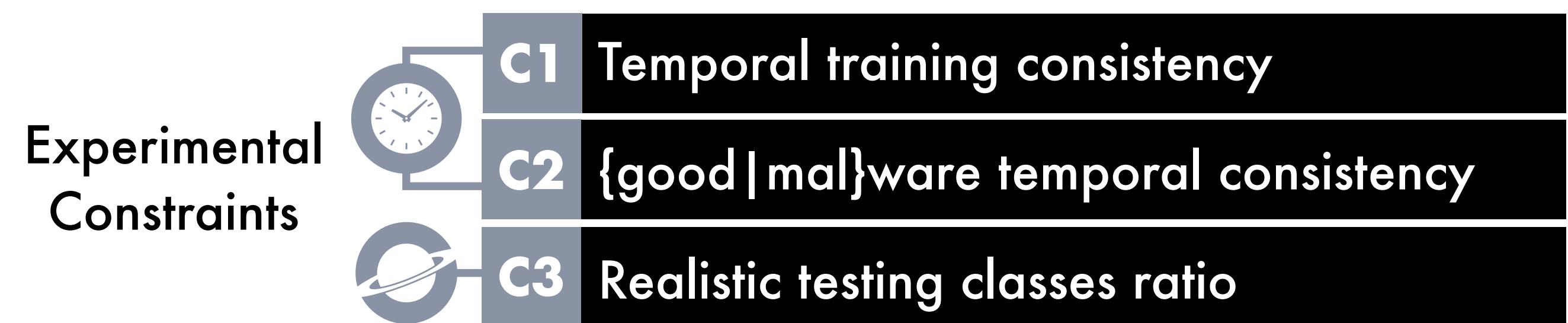


TESSERACT

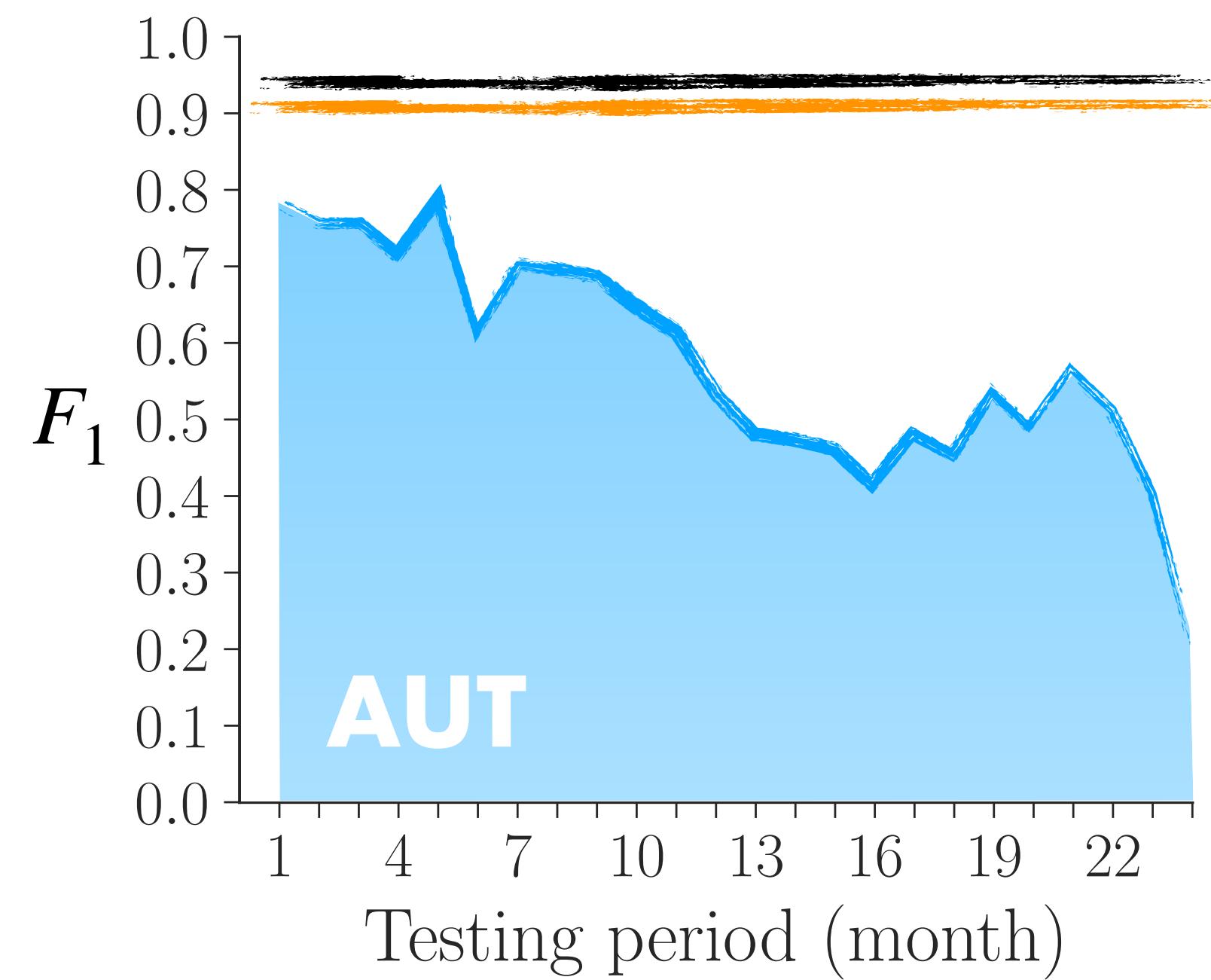
TESSERACT Framework



TESSERACT Framework

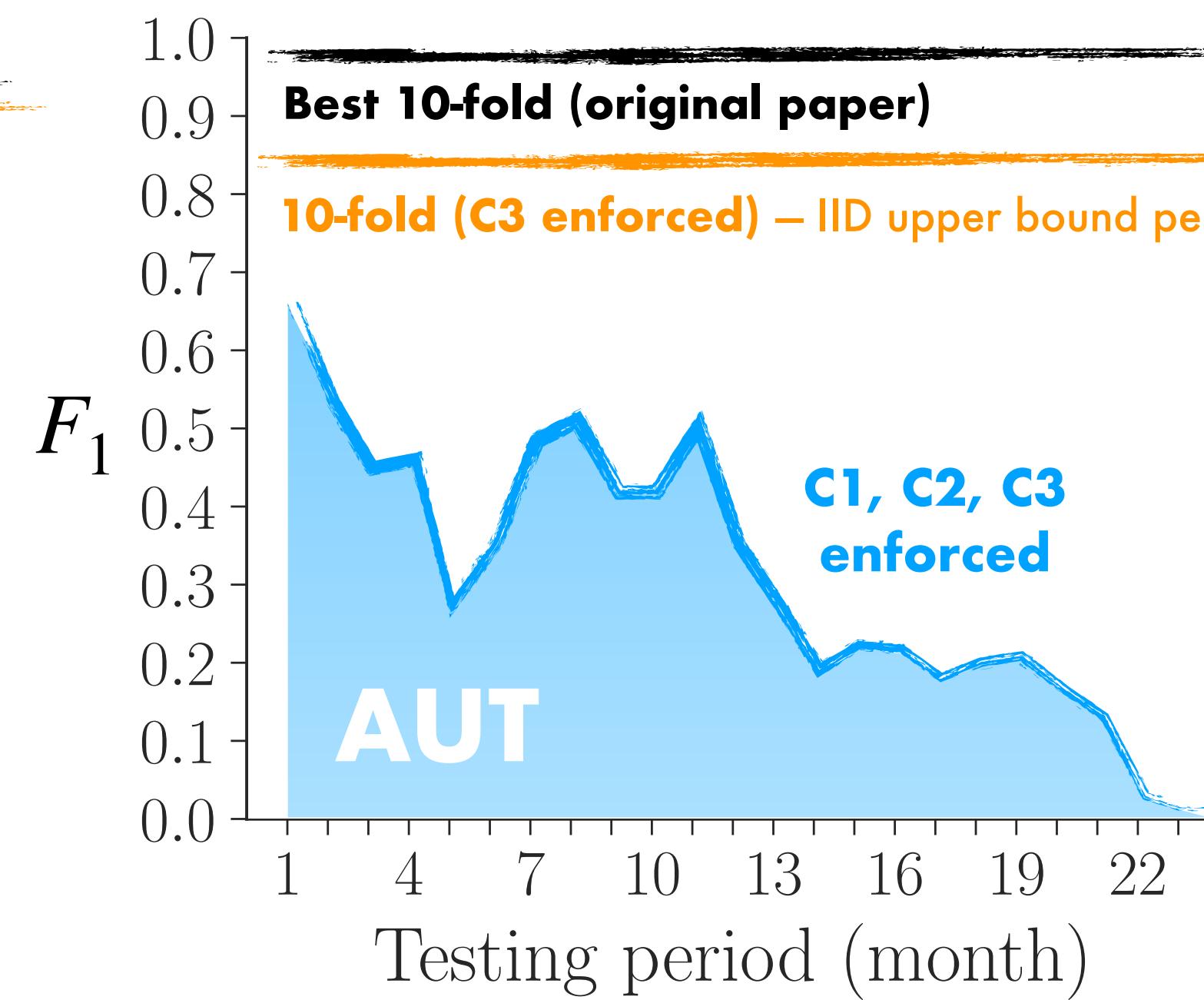
| | | |
|-----------------------------|----|---------------------------------------|
| Experimental Constraints | C1 | Temporal training consistency |
| | C2 | {good mal}ware temporal consistency |
| | C3 | Realistic testing classes ratio |

Method 2



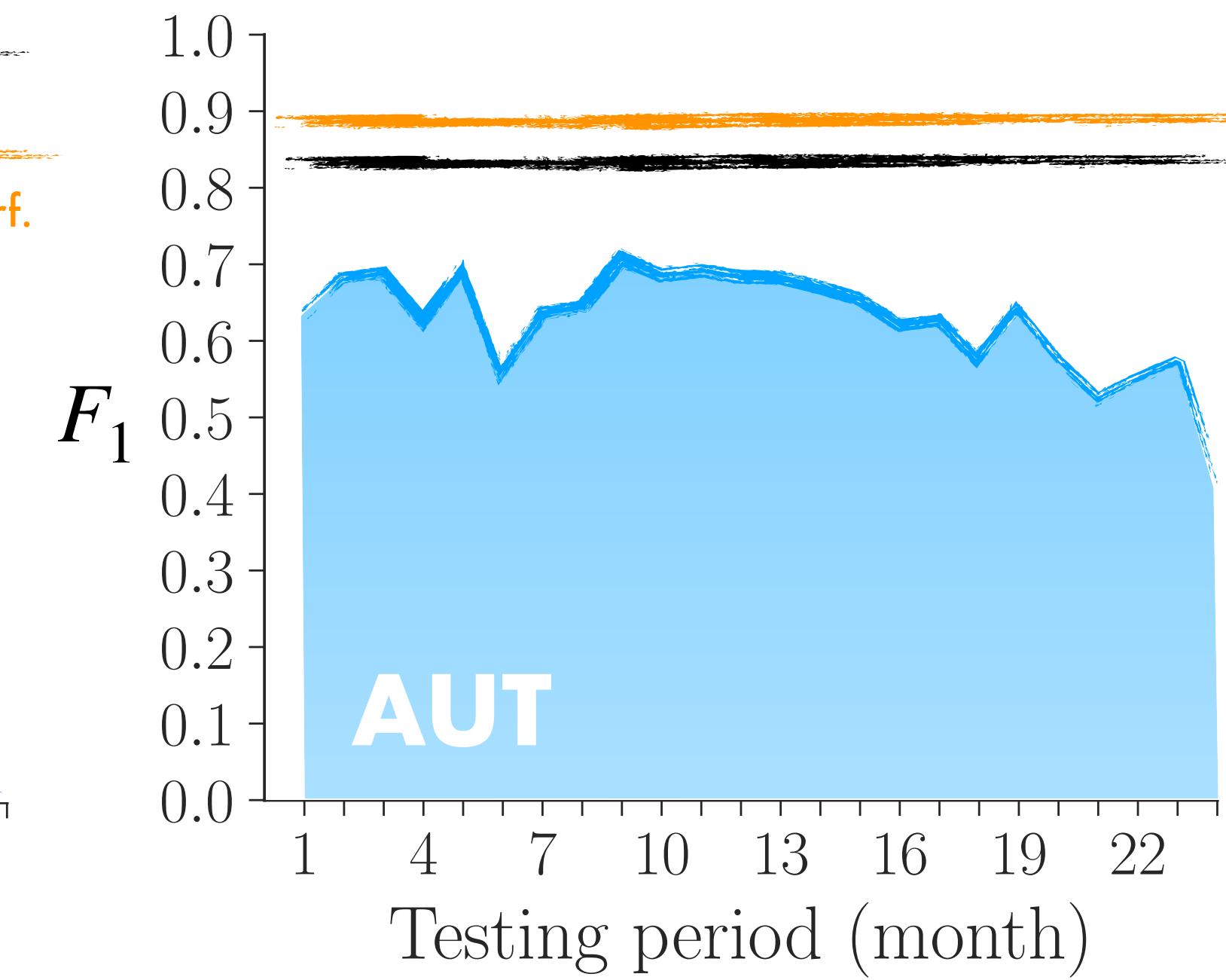
$$AUT(F_1, 24, m) = 0.58$$

Method 1



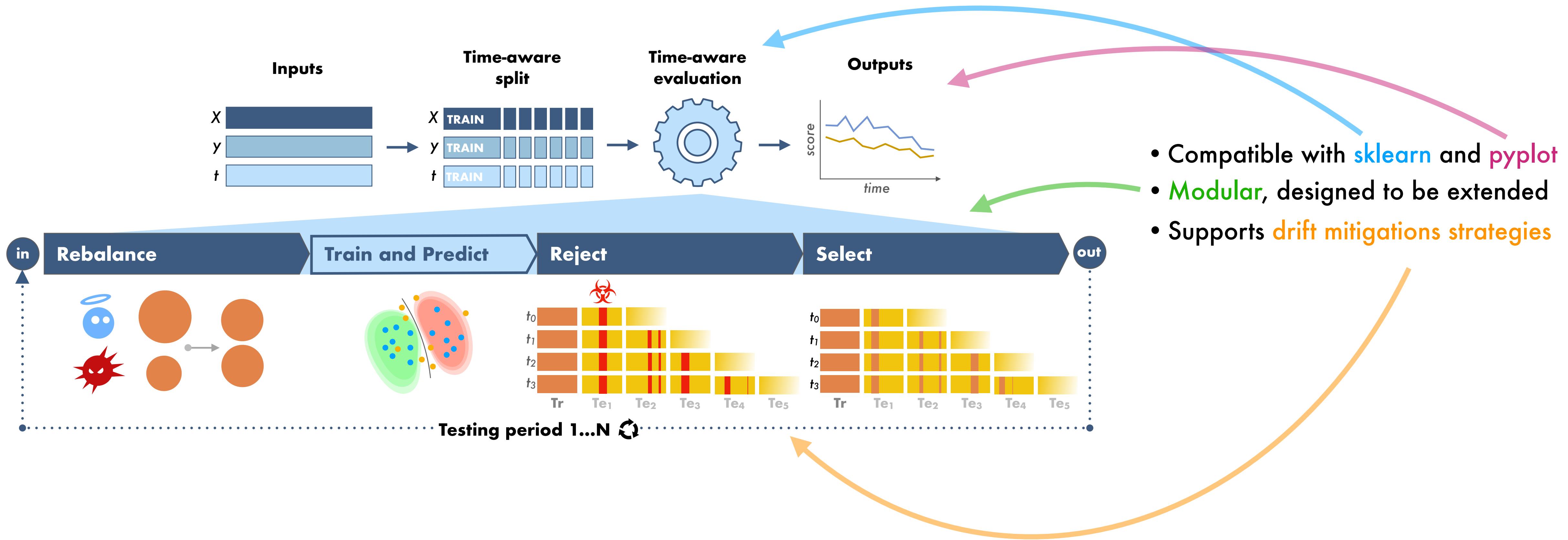
$$AUT(F_1, 24, m) = 0.32$$

Method 3



$$AUT(F_1, 24, m) = 0.64$$

Artifact Design



Artifact: How easy is it to use?

```
1  from sklearn.svm import SVC
2  from tesseract import evaluation, temporal, spatial, temporal, loader, metrics, viz
3
4  def main():
5      # Load features, labels and timestamps
6      X, y, t = loader.load_features('dataset_fname')
7
8      # Partition dataset
9      splits = temporal.time_aware_train_test_split(
10         X, y, t, train_size=12, test_size=1, granularity='month')
11
12     X_train, X_tests, y_train, y_tests, t_train, t_tests = splits
13
14     # Enforce spatio-temporal constraints
15     for y_test, t_test in zip(y_tests, t_tests):
16         temporal.assert_positive_negative_temporal_consistency(y_test, t_test)
17         temporal.assert_train_test_temporal_consistency(t_train, t_test)
18         spatial.assert_class_distribution(y_test, positive_rate=0.1, variance=0.05)
19
20     # Perform a timeline evaluation
21     clf = SVC(kernel='linear', probability=True)
22     results = evaluation.fit_predict_update(clf, *splits)
23
24     # View results
25     metrics.print_metrics(results)
26     # View AUT(F1, 24 months) as a measure of robustness over time
27     print(metrics.aut(results, 'f1'))
28
29     # Generate plot
30     plt = viz.plot_decay(results)
31     plt.show()
32
33
34 if __name__ == '__main__':
35     main()
```

← Imports

← Load timestamped datasets with helpers

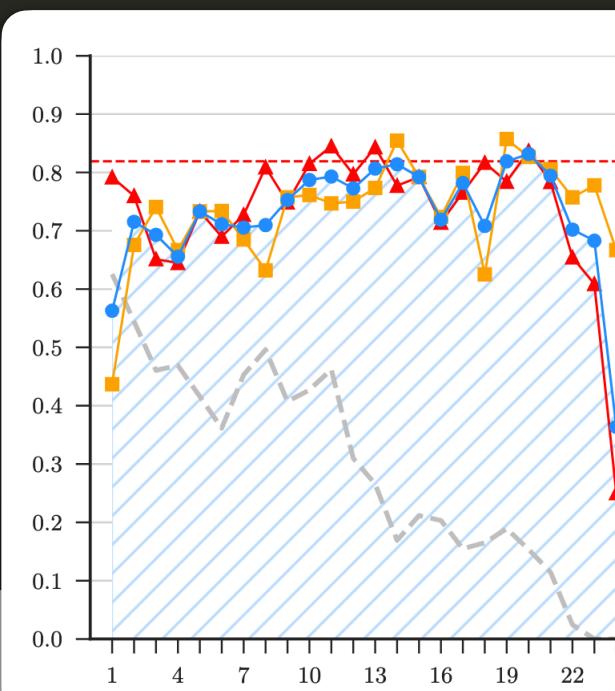
← Split datasets by time in one line

← Assert each constraint to catch accidental bias or when analyzing existing datasets

← Run the evaluation, compatible with `sklearn` classifiers

← View key metrics such as AUT

← Generate and display `pyplot`-compatible charts



Key Influenced Work

Within our team...

- DroidEvolver++ (AISeC '21)
- Transcendent (IEEE S&P '22)
- Drift Forensics (AISeC '23)
- Tesseract (S&P-Magazine '23)

- RPAL (ARTMAN '24)

Through collaboration...

- Insomnia (AISeC '21)
- Do's and Don'ts (USENIX Sec '22)

- Is it Overkill? (DLS '23)

And beyond us...

- APIGraph (ACM CCS '20)
- Continuous Learning (USENIX Sec '23)
- Mateen (RAID '24)
- ... and more!



Academic Impact

450+ citations overall, incl:

- 41 PhD theses
- 12 MSc theses
- 38 Surveys and SoKs

Keynotes at:

- DLS 2023
- RAID 2024



NUS

National University
of Singapore



The Alan Turing Institute



UNIVERSITÉ DU LUXEMBOURG

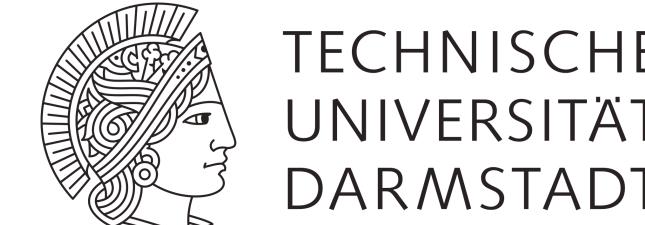
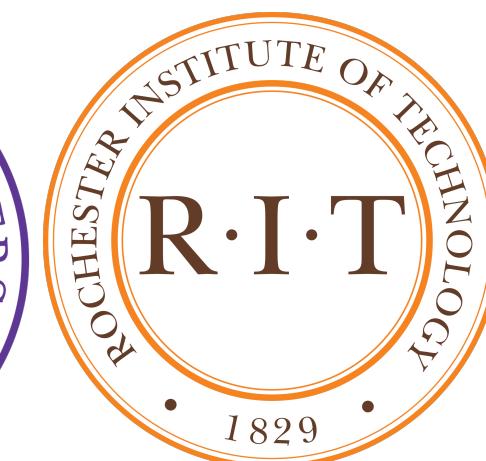
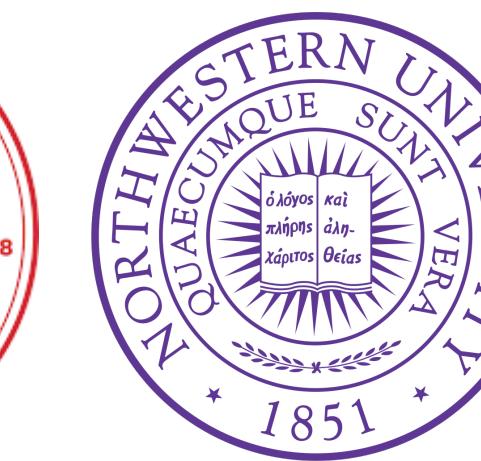


大阪大学
OSAKA UNIVERSITY

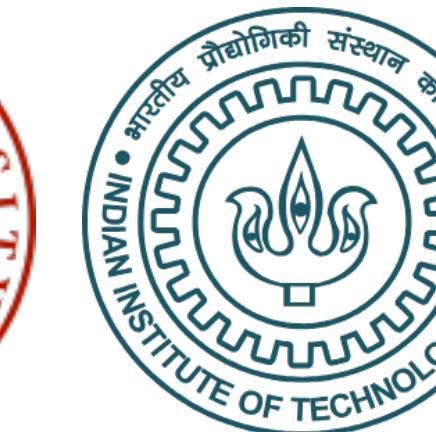
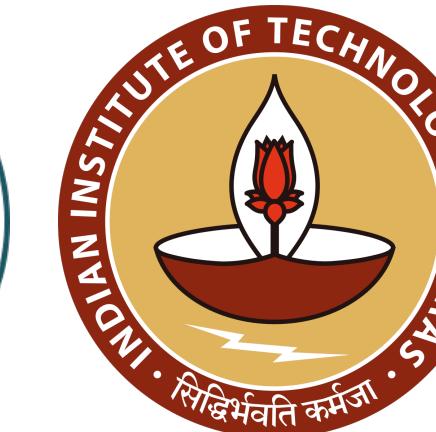


UCL

Artifact shared with **100+** academic institutions across **27** countries, including:



**UNSW
SYDNEY**



Industrial Impact

Artifact shared with **10+** industry and government institutions:



Presented at industry-focused events:

- USENIX ENIGMA, 2019
- Avast CyberSec&AI Connected, 2019
- IBM AI Masterclass, 2024

Educational Impact

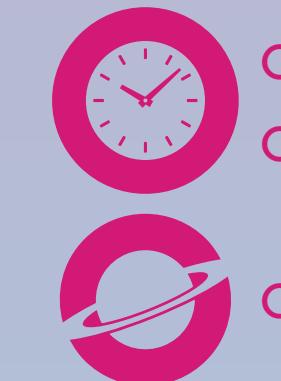
Used as a teaching tool in courses and seminars:

- **University College London**, MSc Information Security course
 - **University of Cagliari**, Ph.D. Course with MLSec labs
 - **University of Bologna**, Ph.D. Course with MLSec labs
 - **University of Modena**, Seminar series
 - **TU Berlin**, Software Engineering PhD and PostDoc Summer School
 - **KU Leuven**, Security & Privacy in Age of AI Summer School
 - **Karlsruhe Institute of Technology**, malware guest lecture series
 - **Imperial College London**, ML and Cyber Security Symposium
 - **Zhejiang University**, Ph.D. course
 - **Trinity College Dublin**, as part of IBM AI Masterclass
- ... and more!

Community Influence

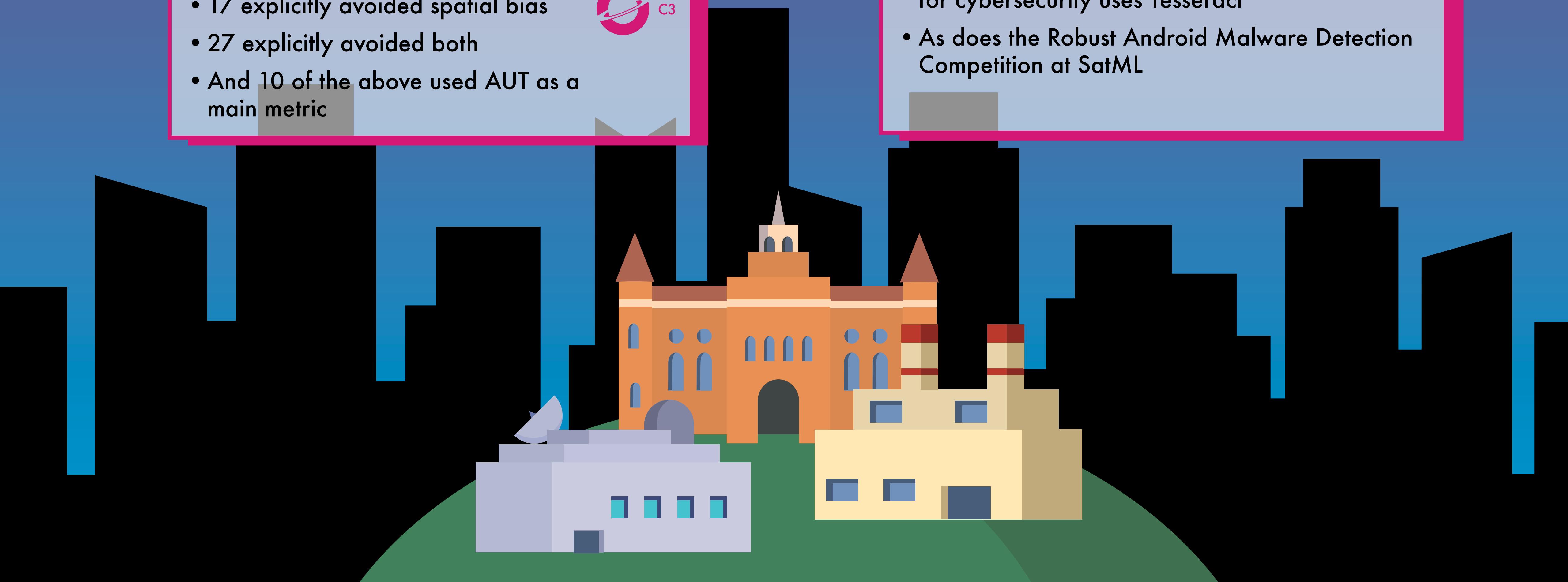
From cited works:

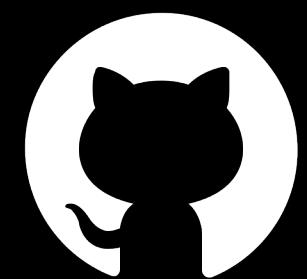
- 34 explicitly avoided temporal bias
- 17 explicitly avoided spatial bias
- 27 explicitly avoided both
- And 10 of the above used AUT as a main metric



Shaping future research...

- Track 3 of the ELSA EU benchmark competition for cybersecurity uses Tesseract
- As does the Robust Android Malware Detection Competition at SatML





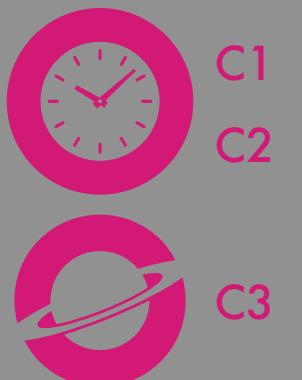
<https://github.com/s2labres/tesseract-ml-release>

“...and releasing the artifact on Github for everyone to use!”



From cited works:

- 34 explicitly avoided temporal bias
- 17 explicitly avoided spatial bias
- 27 explicitly avoided both
- And 10 of the above used AUT as a main metric



- 110+ research teams from 27+ countries
- 75+ papers integrated it in their methodology
- Foundational for future research

<https://github.com/s2labres/tesseract-ml-release>

