NISTIR 8292 DRAFT SUPPLEMENT

Face Analysis Technology Evaluation (FATE)

Part 4: MORPH - Performance of Automated Face Morph Detection

Mei Ngan
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This publication is available free of charge from:
https://www.nist.gov/programs-projects/face-recognition-vendor-test-frvt-ongoing

NATIONAL INSTITUTE OF STANDARDS AND TECHNOLOGY
U.S. DEPARTMENT OF COMMERCE

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U.S. Department of Commerce
Howard Lutnick, Secretary

National Institute of Standards and Technology
Craig Burkhardt, Acting Under Secretary of Commerce for Standards and Technology and Acting NIST Director

FATE MORPH Status and Changelog

Prior editions of this report are maintained on the FATE MORPH website. The FATE MORPH evaluation remains open to new algorithm submissions indefinitely. This report will be updated as new algorithms are evaluated, as new datasets are added, and as new analyses are included. Comments and suggestions should be directed to frvt@nist.gov.

January 9, 2026

- This report adds results for one new algorithm submitted by secunet (secunet-004). See Sections 2.2 and 4.

July 29, 2025

- This report adds results for a new algorithm submitted by Vision-Box (visionbox-001). See Sections 2.2 and 4.

June 6, 2025

- In alignment with the final draft ISO/IEC FDIS 20059 standard on methodologies to evaluate the resistance of biometric recognition systems to morphing attacks, the reporting of BPCER (Bona Fide Presentation Classification Error Rate) in this report has been deprecated and replaced with BSCER (Bona Fide Sample Classification Error Rate).

February 27, 2025

- This report adds results for one new algorithm submitted by secunet (secunet-003). See Sections 2.2 and 4.
- Algorithm timing durations have been added to the FATE MORPH leaderboard.

January 24, 2025

- This report adds results for one new algorithm submitted by the Fraunhofer Institute for Telecommunications Heinrich Hertz Institute (hhi-002). See Sections 2.2 and 4.

December 19, 2024

- This report adds results for a new algorithm submitted by Idemia (idemia-004). See Sections 2.2 and 4.
- We have started reporting error rates at a lower target BSCER of 0.003.

August 14, 2024

- In alignment with the draft ISO/IEC DIS 20059 standard on methodologies to evaluate the resistance of biometric recognition systems to morphing attacks, the reporting of APCER (Attack Presentation Classification Error Rate) in this report has been deprecated and replaced with MACER (Morphing Attack Classification Error Rate).

June 7, 2024

- This report adds results for a new algorithm submitted by Idemia (idemia-003). See Sections 2.2 and 4.

May 10, 2024

- This report adds results for a new algorithm submitted by Vision-Box (visionbox-000). See Sections 2.2 and 4.

April 2, 2024

- This report adds results for a new algorithm submitted by Idemia (idemia-002). See Sections 2.2 and 4.

March 13, 2024

- This report adds results for a new algorithm submitted by Hochschule Darmstadt (hdadfr-006). See Sections 2.2 and 4.

February 21, 2024

- This report adds results for new algorithms submitted by the Norwegian University of Science and Technology (ntnucan-000 and ntnusub-000). See Sections 2.2 and 4.

January 31, 2024

- This report adds results for new algorithms submitted by Idemia (idemia-001) and the Kempelen Institute of Intelligent Technologies (kinit-001). See Sections 2.2 and 4.

October 26, 2023

- This report adds results for one new algorithm submitted by Universidade de Coimbra (visteam-004). See Sections 2.2 and 4.

October 20, 2023

- This report adds results for two new algorithms submitted by the Fraunhofer Institute for Telecommunications Heinrich Hertz Institute (hhi-001) and Neurotechnology (neurotechnology-000). See Sections 2.2 and 4.

June 20, 2023

- This report adds results for two new algorithms submitted by Universidade de Coimbra (visteam-003 and visteamicao-000). See Sections 2.2 and 4.

March 6, 2023

- This report adds results for one new algorithm submitted by West Virginia University (wvudiff-001). See Sections 2.2 and 4.

February 2, 2023

- This report adds results for two new algorithms submitted by Universidade de Coimbra (visteam-002) and University of Bologna (unibo-002). See Sections 2.2 and 4.

December 2, 2022

- This report adds results for one new algorithm submitted by the University of Bologna (unibo-001). See Sections 2.2 and 4.

November 16, 2022

- This report adds results for one new algorithm submitted by secunet (secunet-002). See Sections 2.2 and 4.

September 20, 2022

- This report adds results for two new algorithms submitted by secunet (secunet-001) and Universidade de Coimbra (visteam-001). See Sections 2.2 and 4.

- New results have been published for when additional subject metadata is provided as input to differential morph detection algorithms, which includes sex, age of the subject at the time the probe image is taken, and the age/time difference between the suspected morph and the live probe image. Operationally, this information might be derived from data read from the machine readable zone of a passport for example. These results support measurement of whether algorithms can improve morph detection accuracy when additional subject metadata is provided. See Sections 4.4 and 4.
- Initial results have been published for "morph-resistant" 1:1 face recognition algorithm performance submitted by Universidade de Coimbra (visteam-001). See Sections 4.11, 4.5, and 4.1.3.

July 14, 2022

- This report adds results for one new algorithm (hdamag-001) submitted by Hochschule Darmstadt. See Sections 2.2 and 4.

April 28, 2022

- This report adds results for one new algorithm (wvusingle-002) submitted by West Virginia University. See Section 2.2.

November 29, 2021

- This report adds results for one new algorithm (hdafvdet-001) submitted by Hochschule Darmstadt. See Section 2.2.

October 28, 2021

- This report adds results for two new algorithms submitted by West Virgina University (wvusingle-001) and Universidade de Coimbra (visteam-000). See Section 2.2.
- A new, larger Print + Scanned dataset has been added to the test (and replaces the old Print + Scanned dataset). See Table 6.
- We have retired the Complete, Splicing, and Combined datasets.
- Interactive report cards for each algorithm are published and linked from the accuracy summary table on the FATE MORPH webpage.

September 7, 2021

- This report adds results for one new algorithm (hdafusion-001) submitted by Hochschule Darmstadt. See Section 2.2.

July 27, 2021

- This report adds single-image and differential morph detection results for a new dataset of morphs created using the MIPGAN software provided by the Norwegian University of Science and Technology. See Sections 2.3, 4 and Figure 13.
- An updated version of the FATE MORPH API document has been published. This update adds an optional input parameter to the function detectMorphDifferentially(). The additional parameter represents the time/age difference (in days) between a suspected morph and the live probe image.
- Interactive report cards for each algorithm have been published and linked from the results table on the FATE MORPH webpage. For example, `https://pages.nist.gov/frvt/reportcards/morph/hdalbp_005_1.html`.

April 16, 2021

- This report adds algorithm score distribution plots (Section **??**) and APCER calibration plots (Section 4.8).
- This report updates the differential bona fide morph detection scores vs. elapsed time plots, now with results for both visa and mugshot bona fide datasets. See Section 4.9.
- Interactive report cards for each algorithm will be published and linked from the FATE MORPH webpage in the coming weeks.

February 02, 2021

- This report updates the morph detection error metrics, attack presentation classification error rate (APCER) and bona fide sample classification error rate (BSCER), to incorporate when an algorithm fails to process an image. See Sections 3 and 4.1.2. Results in all tables and plots in this report and on our website reflect this change unless otherwise noted.
- This report includes single-image and differential morph detection results for three new datasets of morphs (Visa-Border, UNIBO Automatic Morphed Face Generation Tool v2.0, Twente) created using new and updated methods provided by the University of Twente and the University of Bologna. See Sections 2.3 and 4.
- This report adds differential morph detection results for the UNIBO Automatic Morphed Face Generation Tool v1.0 dataset. See Sections 2.3, 4.3, and Figure 9.
- This report replaces the term ”match score” with ”comparison score” where applicable to better align with standard terminology.

July 24, 2020

- This report adds results for one new algorithm (hdadfr-003) submitted by Hochschule Darmstadt. See Section 2.2.

June 3, 2020

- This report adds results for two new algorithms (hdadfr-002, hdalaplace-001) submitted by Hochschule Darmstadt. See Section 2.2.
- This report adds a new dataset to support assessment of image resolution on morph detection accuracy. See Section 2.3.
- This report documents initial analyses on the impact of image resolution on single-image morph detection accuracy. See Executive Summary and Section 4.7.

March 4, 2020

- This report has been formally published ast NIST Interagency Report (NISTIR) 8292.

January 24, 2020

- This report adds results for seven new algorithms submitted by Hochschule Darmstadt and one new algorithm submitted by the Norwegian University of Science and Technology. See Section 2.2.
- This report includes results for a new dataset of morphs provided by the University of Lincoln. See Section 4.6.3.
- This report includes results for a new dataset of bona fide images, which includes 1) a set of high quality visa portraits for single-image morph detection and 2) a set of high quality visa portraits + a set webcam probes that exhibit moderately poor pose variations and background illumination for two-image differential morph detection. See Sections 2.3, 4.1.1, 4.2, and 4.3.

- Sample imagery for the new datasets have been added to Figures 2 and 3.
- The accuracy results in Tables 4.2 and 4.3 are now grouped by dataset and ordered by algorithm accuracy (MACER @ BSCER_m=0.01).
- This report documents new analyses, including 1) BSCER as a function of morph detection score threshold across visa and mugshot datasets and 2) for two-image differential morph detection, bona fide morph detection score as a function of time elapsed between the bona fide and probe image.
- We have migrated our website to a new platform that supports interactive plotting and sortable tables: https://pages.nist.gov/frvt/html/frvt_morph.html. Summary accuracy tables and DET plots are published on the website and will be updated as new results are available.

September 17, 2019

- This is the first FATE MORPH report published as a draft for public comment. This report documents results for five morph detection algorithms over twelve datasets.

Executive Summary

Background

Face morphing and the ability to detect it is an area of high interest to photo-credential issuance agencies, companies, and organizations employing face recognition for identity verification. Face morphing is an image manipulation technique where two or more subjects' faces are morphed or blended together to form a single face in a photograph. Morphed photos can look very realistically like all contributing subjects. Morphing is easy to do and requires little to no technical experience given the vast availability of tools available at little or no cost on the internet and mobile platforms. If a morphed photo gets onto an identity credential for example, multiple, if not all constituents of the morph, can use the same identity credential. Morphs can be used to fool both humans [1] [2] and current face recognition systems [3], which presents a vulnerability to current identity verification processes.

FATE MORPH Test Activity

The FATE MORPH test provides ongoing independent testing of prototype face morphing attack detection (MAD) technologies. The evaluation is designed to obtain commonly measured assessment of morph detection capability to inform developers and end-users. FATE MORPH is open for ongoing participation worldwide, and there is no charge to participate. The test opened in June 2018, and NIST has since received a number of morph detection algorithm submissions from international academic and commercial entities.

The test leverages a number of datasets created using different morphing methods with goals to evaluate algorithm performance over a large spectrum of morphing techniques. Testing was conducted using a tiered approach, where algorithms were evaluated on low quality morphs created with readily accessible tools available to non-experts, morphs generated using automated morphing methods based on academic research, and high quality morphs created using commercial-grade tools. We'd like to get an assessment on the existence and extent of morph detection capabilities, and if there is indication of high accuracy, much larger datasets can be curated to support large-scale evaluation of the technology.

Results and Notable Observations

To assess morph detection performance, two primary quantities are reported - the Morphing Attack Classification Error Rate (MACER) or morph miss rate and the Bona Fide Sample Classification Error Rate (BSCER) or false detection rate. MACER and BSCER are reported both individually and as a tradeoff in the DET analysis in this report. *Section 3*

Ideally, it is important that morph detection technology produce very low false detection rates given the assumption that most transactions will be on legitimate photos that are not morphs. False detection rates need to be controlled, because additional amounts of resources will be required to adjudicate such errors. With that said, an initial automated morph detection capability with say ideally 0% false detection rates but high morph miss rates would still yield gains in operations compared to not having any morph detection capability at all.

- **Single-image Morph Detection:** In this use case, a single image is provided to the algorithm, and the software has to 1) make a decision on whether it thinks the image is a morph and 2) provide a confidence score on its decision.

  For some recent algorithms (e.g., idemia-004), we observe reduced morph miss rates at a false detection rate of 0.003, particularly on a number of tier 1 (low quality) and tier 2 (automated) datasets. While recent progress has been observed in single-image morph detection, many of the algorithms do not generalize well across different unseen morphing methods, and error rates remain high on tier 3 (high quality) datasets, which is indicative that morph detection with a single image in isolation remains a challenging research issue. *Section 4.2, 4.6*

  Caveat: There is an exception to the generally high morph miss rates observed, which is the University of Bologna's algorithm (unibo-000) result against morphs created using techniques developed also by the University of Bologna in the UNIBO Automatic Morphed Face Generation Tool v1.0 and v2.0 datasets. Those particular datasets were generated using a set of sequestered source images and morphed using software that implemented techniques published in [3–6]. The unibo-000 algorithm's morph miss rate is 0.09 and 0.16 at a false detection rate of 0.01 on datasets generated with their v1.0 and v2.0 tool respectively. While such results need to be caveated, it highlights an interesting data point which quantifies that morph detection software can be trained/designed to detect images created

using a particular morphing process and confirms the importance of cross-database training and testing for the development and evaluation of morphing detection algorithms. *Section 4.2.2*

Image Resolution: We conducted an initial study on whether image resolution has an impact on single-image morph detection accuracy. The results show that some algorithms are able to take advantage of additional resolution in images and reduced error rates are observed as image resolution increases. For those algorithms, there appears to be diminishing returns in error reduction when the interocular distance (IOD) is larger than 600 pixels. These results are caveated with necessary assessments of MACER (morph miss rates) and BSCER (false detection rates) separately as a function of score threshold. Interestingly, we observe that while false detection rates decrease in higher resolution images (at equal thresholds), morph miss rates increase as resolution increases (at equal thresholds).

The implications of these initial results would mean for ecosystems that only expect and can enforce processing of images at high resolution, then the use of higher resolution photos would yield reductions in error rates, for some algorithms. But, consequently, in a morph detection system that is set to a threshold configured for higher resolution photos, if it encounters lower resolution photos, the system would expect 1) increased false detection rates but favorably, 2) decreased morph miss rates. Likewise, in a system that is configured at a threshold targeted for lower resolutions, when higher resolution photos are encountered, the system would observe, favorably, decreased false detection rates, but unfavorably, increased morph miss rates. The existence and magnitude of these observations vary between algorithms. *Section 4.7*

- **Two-image Differential Morph Detection:** In this use case, two face photos are provided to the algorithm, the first being a suspected morph and the second image representing a known, non-morphed face image of one of the subjects contributing to the morph (e.g., live capture image from an eGate). The software has to 1) make a decision on whether it thinks the image is a morph and 2) provide a confidence score on its decision. This procedure supports measurement of whether algorithms can detect morphed images when additional information (the second photo) is provided.

  Recent developments in differential morph detection based on the approach published in [7] has shown promising results, as demonstrated by the most recent algorithm submitted by secunet (secunet-002) where morph miss rates ranging from 9% to 36% across all datasets can be achieved at a false detection rate of 0.01 (1 in 100). At a false detection rate of 0.02 (2 in 100), the morph miss rates are reduced to between 4% to 22% across all datasets evaluated. This is a step forward in morph detection capability and existence proof that morph detection with the presence of an additional live probe image (e.g., e-gate authentication) may be a viable and generalizable approach. With that said, there are additional reductions in error rates to be made to advance the state of the art to target even lower false detection rates, given morph miss rates are currently 100% at a more stringent false detection rate of 0.001 (1 in 1000).

  One possible reason for better generalizability observed in the differential morph detectors is that some of the algorithms are using identity information derived between the image and live probe photo for morph determination, rather than detection of particular morphing artifacts that may differ across morphing methods. For the set of hdaarcface and hdadfr algorithms, we observe elevated false detection rates (BSCER) due to ageing effects. As the time elapsed between a bona fide image and the live probe image increases, the occurrence of bona fides being incorrectly classified as morphs also increases, indicating that the differential morph detectors have difficulty deconflicting changes in appearance due to ageing (and incorrectly flagging legitimate photos as being a morph). The secunet-001 and secunet-002 algorithms (more recent enhancements of the hdaarcface [7] approach) do not exhibit such ageing-driven increase in false detection rates and achieve lower error rates than its predecessors.

- **Printing and Scanning:** The process of printing and scanning (printing a digital image onto paper, then scanning it back in) or re-digitalization is known to be one of the biggest challenges to morph detection. The process of printing and scanning photos is followed by a number of identity credential issuance entities (e.g. passports) worldwide in countries that rely on mail-in applications. Therefore, the use case of morph detection on printed and scanned photos is very relevant. We investigate the performance of algorithms on print and scanned photos using a subset of visa-like images (both morphs and nonmorphs) from a global population, with live digital probe images of border

crossing photographs collected with a webcam (Table 6). Algorithm behavior varies between different morph detection methods – many differential morph detectors, at a developer-defined threshold, show low morph miss rates BUT very high false detection rates, which means the algorithms are classifying most scanned photos as morphs, even when they're not. Some single-image morph detectors show very low false detection rates BUT very high morph miss rates, which could be indicative of reduction or elimination of morphing artifacts during the print-and-scan process. Nevertheless, error rates on print and scanned photos remain high at operationally-relevant false detection rates. *Section 4.2.3*.

We continue to expand our test to evaluate differential morph detection capabilities across a spectrum of morphing methods and types of imagery. *Section 4.3*

Future Work

FATE MORPH will run continuously, and this report will be updated as new algorithms, datasets, analyses, and metrics are added.

Acknowledgements

The authors would like to thank the Department of Homeland Security's Science and Technology Directorate and Office of Biometric Identity Management, U.S. Department of State, Federal Bureau of Investigation, Noblis, MITRE, Otto von Guericke University of Magdeburg, University of Bologna, Australian Defence Science and Technology Group, University of Lincoln, University of Twente, and the Norwegian University of Science and Technology for their collaboration and contributions to this activity. Additionally, the authors are grateful to Hochschule Darmstadt for discussions on test methodology and metrics.

The authors are grateful to staff in the NIST Biometrics Research Laboratory for infrastructure supporting rapid evaluation of algorithms.

Disclaimer

Specific hardware and software products identified in this report were used in order to perform the evaluations described in this document. In no case does identification of any commercial product, trade name, or vendor, imply recommendation or endorsement by the National Institute of Standards and Technology, nor does it imply that the products and equipment identified are necessarily the best available for the purpose.

The data, protocols, and metrics employed in this evaluation were chosen to support morph detection research and should not be construed as indicating how well these systems would perform in applications. While changes in the data domain, or changes in the amount of data used to build a system, can greatly influence system performance, changing the task protocols could reveal different performance strengths and weaknesses for these same systems.

Institutional Review Board

The National Institute of Standards and Technology's Research Protections Office reviewed the protocol for this project and determined it is not human subjects research as defined in Department of Commerce Regulations, 15 CFR 27, also known as the Common Rule for the Protection of Human Subjects (45 CFR 46, Subpart A).

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1 The FATE MORPH Activity

Face morphing and the ability to detect it is an area of high interest to a number of photo-credential issuance agencies and those employing face recognition for identity verification. Face morphing is an image manipulation technique where two or more subjects' faces are morphed or blended together to form a single face in a photograph. Morphed photos can look very realistically like all contributing subjects. If a morphed photo gets onto an identity credential for example, multiple, if not all constituents of the morph, can use the same identity credential. Morphs can be used to fool both humans [1] [2] and current face recognition systems [3], which presents a vulnerability to current identity verification processes. Figure 1 illustrates the impact of morphed photos on current algorithms from some of the leading face recognition algorithms (labeled as A, B, C, and D) submitted to the NIST Ongoing FRTE 1:1 Verification test. The overlap between the morph and genuine comparison score distributions, and the significant percentage of morph comparisons that would successfully authenticate at FMR=0.001 (1 in 1000) provides the basis for research into how to detect this form of image manipulation.

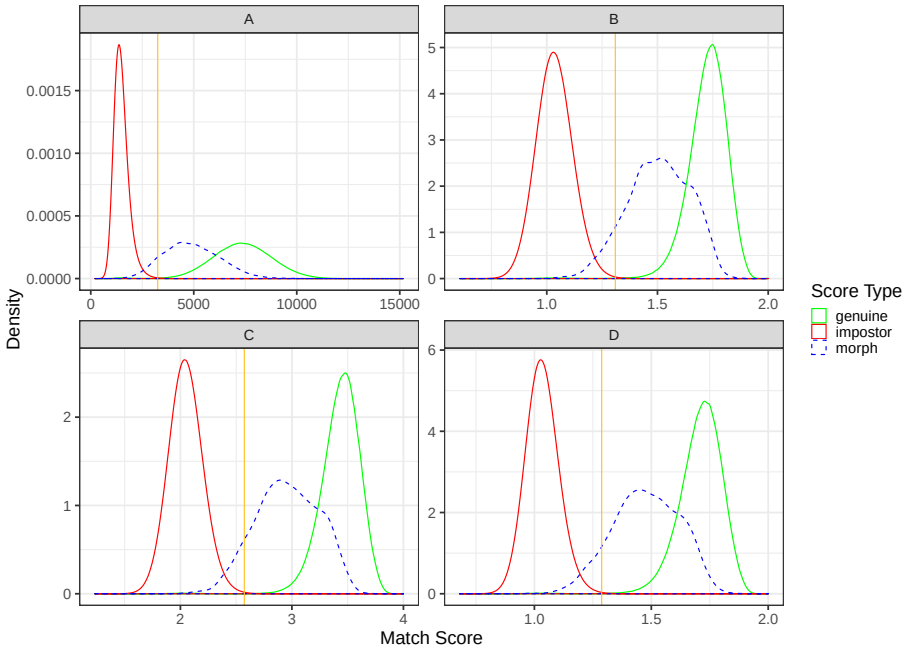


Figure 1: Morph comparison score distribution. The plot shows comparison score distribution for 1) genuine comparisons of photos of the same person (green) 2) imposter comparisons of photos of different people (red), and 3) morph comparisons of morphed photos with other photos of contributing subjects (blue). The gold line represents the score threshold at a false match rate (FMR) of 0.001. All comparison scores to the right of the gold line indicates that the algorithm thinks the photos are of the same person at that FMR threshold (e.g. successful authentication at an eGate).

The FATE MORPH test will provide ongoing independent testing and measurement of prototype face morph detection technologies. The evaluation is designed to obtain an assessment of morph detection capability to inform developers and end-users, and will evaluate two separate tasks:

- Algorithmic capability to detect face morphing (morphed/blended faces) in still photographs:
  - Single-image morph detection of non-scanned photos, printed-and-scanned photos, and images of unknown photo format/origin;
  - Two-image differential morph detection of non-scanned photos, printed-and-scanned photos, and images of unknown photo format/origin. This procedure supports measurement of whether algorithms can detect mor-

phed images when an additional live capture image or when additional information about the subject is provided.

- Face recognition algorithm resistance against morphing. The expected behavior from algorithms is to be able to correctly reject comparisons of morphed images against all constituents that contributed to the morph. The goal is to show algorithm robustness against morphing alterations when morphed images are compared against other images of the subjects used for morphing.

2 Methodology

2.1 Test Environment

The evaluation was conducted offline at a NIST facility. Offline evaluations are attractive because they allow uniform, fair, repeatable, and large-scale statistically robust testing. Testing was performed on high-end server-class blades running the CentOS Linux [8] operating system. The test harness used concurrent processing to distribute workload across dozens of computers.

2.2 Algorithms

The FATE MORPH program is open to participation worldwide. The participation window opened in June 2018, and the test will evaluate algorithms on an ongoing basis. There is no charge to participate. The process and format of algorithm submissions to NIST are described in the FATE MORPH Concept, Evaluation Plan, and Application Programming Interface (API) document [9]. Participants provide their submissions in the form of libraries compiled on a specified Linux kernel, which are linked against NIST's test harness to produce executables. NIST provides a validation package to participants to ensure that NIST's execution of submitted libraries produces the expected output on NIST's test machines.

This report documents the results of all algorithms submitted for testing to date. Tables 1, 2, and 3 list the participants who submitted algorithms to FATE MORPH.

Participant Name	Short Name	Submission Sequence	Submission Date	Developer Notes
University of Bologna	unibo	000	2019.07.29	
Norwegian University of Science and Technology	ntnussl	001 002	2019.07.08 2019.10.11	[10]
Hochschule Darmstadt	hdalbp	006	2019.12.02	The idea behind the LBP implementation is based on HDA (http://dasec.h-da.de) / NTNU (https://www.ntnu.edu/nbl) approaches and published in [11–13].
Hochschule Darmstadt	hdabsif	004	2020.01.17	
Hochschule Darmstadt	hdaprnu	004	2020.01.21	The idea behind the PRNU implementation is based on a HDA (http://dasec.h-da.de) / PLUS (http://www.wavelab.at) cooperation and published in [14, 15].

Hochschule Darmstadt	hdalaplace	001	2020.04.01	
Hochschule Darmstadt	hdafusion	001	2021.08.24	The idea behind the hdafusion implementation will be published in [16].
Hochschule Darmstadt	hdafvdet	001	2021.11.05	
West Virginia University	wvusingle	001 002	2021.09.10 2022.04.21	The idea behind the wvusingle implementation is published in [17].
Universidade de Coimbra	visteam	000 001 002 003 004	2021.10.12 2022.08.03 2023.01.05 2023.05.23 2023.10.20	The idea behind the visteam implementation is published in [18].
Universidade de Coimbra	visteamicao	000	2023.05.23	
Fraunhofer Institute for Telecommunications Heinrich Hertz Institute	hhi	001	2023.10.04	The hhi implementation involves a feature focus detector [19] in HSV color space trained on differently generated morphed face images, e.g., Style-Transfer improved morphs [20].
Neurotechnology	neurotechnology	000	2023.10.13	
Vision-Box	visionbox	000	2024.03.25	
Idemia	idemia	003	2024.05.18	
Idemia	idemia	004	2024.12.16	
Fraunhofer Institute for Telecommunications Heinrich Hertz Institute	hhi	002	2025.01.17	
Vision-Box	visionbox	001	2025.07.14	

Table 1: FRVT MORPH Participants (Single-image Morph Detection)

Participant Name	Short Name	Submission Sequence	Submission Date	Developer Notes
Hochschule Darmstadt	hdawl	002	2019.12.02	The hdawl submission is a weighted landmark analysis approach (i.e., difference of landmarks) and is based on the work described in [21, 22].
Hochschule Darmstadt	hdalbp	006	2019.12.02	The idea behind the LBP implementation is based on HDA (http://dasec.h-da.de) / NTNU (https://www.ntnu.edu/nbl) approaches and published in [11–13].
Hochschule Darmstadt	hdabsif	004	2020.01.17	
Hochschule Darmstadt	hdalaplace	001	2020.04.01	

Hochschule Darmstadt	hdafusion	001	2021.08.24	The idea behind the hdafusion implementation is published in [16].
Hochschule Darmstadt	hdamag	001	2022.07.06	
Universidade de Coimbra	visteam	000 001 002 003 004	2021.10.12 2022.08.03 2023.01.05 2023.05.23 2023.10.20	The idea behind the visteam implementation is published in [18].
secunet	secunet	001 002	2022.08.11 2022.11.08	The secunet implementation is an enhancement of the approach described in [7].
University of Bologna	unibo	001	2022.11.17	
West Virginia University	wvudiff	001	2023.03.02	The idea behind the WVU implementation is partially based on [23].
Universidade de Coimbra	visteamicao	000	2023.05.23	
Kempelen Institute of Intelligent Technologies	kinit	001	2024.01.24	
Norwegian University of Science and Technology	ntnucan	000	2024.02.13	The idea behind the ntnu implementation is partially based on [24,25].
Norwegian University of Science and Technology	ntnusub	000	2024.02.13	The idea behind the ntnu implementation is partially based on [24,25].
Hochschule Darmstadt	hdaarcface hdadfr hdadfr hdadfr	001 002 003 006	2019.12.29 2020.04.01 2020.07.15 2024.03.02	The idea behind the hdaarcface/hdadfr implementation is published in [7].
Idemia	idemia	001 002	2024.01.25 2024.03.20	
Vision-Box	visionbox	000	2024.03.25	
Idemia	idemia	003	2024.05.18	
secunet	secunet	003	2025.02.21	.
Vision-Box	visionbox	001	2025.07.14	
secunet	secunet	004	2025.12.19	.

Table 2: FRVT MORPH Participants (Two-image Differential Morph Detection)

Participant Name	Short Name	Submission Sequence	Submission Date	Developer Notes

Universidade de Coimbra	visteam	000 001 002 003 004	2021.10.12 2022.08.03 2023.01.05 2023.05.23 2023.10.20	
Universidade de Coimbra	visteamicao	000	2023.05.23	
Vision-Box	visionbox	000	2024.03.25	
Vision-Box	visionbox	001	2025.07.14	

Table 3: FRVT MORPH Participants (Morph Resistant 1:1 Face Recognition)

2.3 Image Datasets

Testing was performed over a number of datasets created using various methods with goals to evaluate algorithm performance over a large spectrum of morphing techniques. Testing was conducted using a tiered approach, where algorithms were evaluated on

- **Tier 1:** Lower quality morphs created with readily accessible tools available to non-experts, such as online tools from public websites and free mobile applications. These morphs are created using low effort processes and are generally low quality and contain large amounts of morphing artifacts that are visible to the human eye.
- **Tier 2:** Morphs generated using automated morphing methods based on academic research and best practices. Automated methods allow for generation of morphs in large quantities for testing.
- **Tier 3:** Higher quality morphs created using either commercial-grade tools with manual processes or generated with automated methods and manually post-processed to remove artifacts. These are high quality morphs with very minimal visible morphing artifacts.

All source images used to generate the morphs in the test datasets are frontal, portrait-style photos. Dataset information is summarized in Tables 4, 5, 6, and sample imagery is provided in Figure 2. For morph detection, each image is accompanied by an associated image label describing the image format/origin, which includes non-scanned photos, printed-and-scanned photos, and photos of unknown format.

- **Non-scanned photos:** Photos are digital images known to not have been printed and scanned from paper. There are a number of operational use-cases for morph detection on such digital images.
- **Printed-and-scanned photos:** While there are existing techniques to detect manipulation of a digital image, once the image has been printed and scanned from paper, it leaves virtually no traces of the original image ever being manipulated. So the ability to detect whether a printed-and-scanned image contains a morph warrants investigation.
- **Photos of unknown format:** In some cases, the format and/or origin of the image in question is not known, so images with "unknown" labels will also be tested.

2.3.1 Tier 1 - Low Quality Morphs

Dataset	Morphing Method	# Morphs	# Source Images	Image Size	Notes
Online tool from website	Unknown	1183	558	300x400	The probe images used to evaluate differential MAD on this dataset are portrait quality images.
Global Morph	Automated	1346	254	512x768	Entire source images are averaged after alignment and feature warping. Morphs were created using subjects of the same sex and ethnicity labels. The probe images used to evaluate differential MAD on this dataset are portrait quality images.

Table 4: Tier 1 datasets: morphs created with easily accessible, non-expert morphing software such as online tools from websites and mobile applications. All morphs are created with two subjects and subject alpha, where known, is 0.5 (i.e., each subject contributed equally to the morph). The image label represents the label that was provided to the algorithm while processing images from the particular dataset.

2.3.2 Tier 2 - Automated Morphs

Dataset	Morphing Method	# Morphs	# Source Images	Image Size	Notes
Local Morph	Automated	1346	254	512x768	Only the face area is averaged after alignment and feature warping; Subject A provides the periphery. Morphs were created using subjects of the same sex and ethnicity labels. The probe images used to evaluate differential MAD on this dataset are portrait quality images.
Local Morph Colorized Average	Automated	1346	254	512x768	Only the face area is averaged after alignment and feature warping. Subject A provides the periphery. Face area is adjusted to the average of Subject A's and Subject B's face color histograms. Morphs were created using subjects of the same sex and ethnicity labels. The probe images used to evaluate differential MAD on this dataset are portrait quality images.

Local Morph Colorized Match	Automated	1346	254	512x768	Only the face area is averaged after alignment and feature warping. Subject A provides the periphery. Face area is adjusted to match Subject A's color histogram. Morphs were created using subjects of the same sex and ethnicity labels. The probe images used to evaluate differential MAD on this dataset are portrait quality images.
UNIBO Automatic Morphed Face Generation Tool v1.0 [3–5]	Automated	2464	64	median: 696x928, min: 488x651, max: 788x1051	Morphs were created using subjects of the same sex and ethnicity labels. The probe images used to evaluate differential MAD on this dataset are informal photos, often with pose angle and illumination variations. These photos were often collected with a webcam and the subject looking at the camera.
Image Resolution	Automated	19978 per image resolution	251 per image resolution	Median: 4612x6149 (1200 IOD), 2306x3075 (600 IOD), 577x769 (300 IOD), 289x385 (150 IOD), 145x193 (75 IOD)	Morphs were created using the UNIBO Automatic Morphed Face Generation Tool v2.0 [3–6] at the highest resolution (1200 IOD), then resized to lower resolutions. Morphs were created using subjects of the same sex and ethnicity labels.
Visa-Border	Automated	25727	51454		Morphs were created using the UNIBO Automatic Morphed Face Generation Tool v2.0 [3–6]. Morphs were created using subjects of similar age and with the same sex and nationality labels. Source images used for morphing are visa-like images from a global population, and the live probe images are border crossing photographs collected with a webcam of travelers entering the United States. The border crossing photos often have pose angle and illumination variations.

UNIBO Automatic Morphed Face Generation Tool v2.0 [3–6]	Automated	2464	64		Morphs were created using subjects of the same sex and ethnicity labels. The probe images used to evaluate differential MAD on this dataset are informal photos, often with pose angle and illumination variations. These photos were often collected with a webcam and the subject looking at the camera.
Twente	Automated	2464	64		Face landmarks are detected based on [26], and automatic post-processing/splicing is based on [27]. Morphs were created using subjects of the same sex and ethnicity labels. The probe images used to evaluate differential MAD on this dataset are informal photos, often with pose angle and illumination variations. These photos were often collected with a webcam and the subject looking at the camera.
MIPGAN-II [28, 29]	Automated	2464	64		Morphs were created using subjects of the same sex and ethnicity labels. The pre-trained network models were fine-tuned on the source imagery used to generate the morphs. The probe images used to evaluate differential MAD on this dataset are informal photos, often with pose angle and illumination variations. These photos were often collected with a webcam and the subject looking at the camera.

Table 5: Tier 2 datasets: morphs created using various automated methods. All morphs are created with two subjects and subject alpha, where known, is 0.5 (i.e., each subject contributed equally to the morph). The image label represents the label that was provided to the algorithm while processing images from the particular dataset.

2.3.3 Tier 3 - High Quality Morphs

Dataset	Morphing Method	# Morphs	# Source Images	Image Size	Notes
Manual	Commercial Tools	323	825	640x640, 1080x1080	The probe images used to evaluate differential MAD on this dataset are portrait quality images.
Lincoln [30]	Automated + Manual	108	-	445x580	

DST	Automated	171	487	1350x1350, 900x1200, 512x768	Subject A provides the periphery. Faces are detected using the Viola-Jones [31] algorithm. [32] is applied to establish initial facial landmark points, with additional landmark points synthesized as necessary. Techniques including Delaunay triangulation are used to develop warpable meshes, which are rendered using affine warping. For minimization of morphing artifacts, denoising and sharpening methods are applied. Morphs were created using subjects of the same sex and ethnicity labels.
Print + Scanned		3604	2739	600x600	A subset of the morphs and bona fides from the Visa-Border dataset were printed on photo paper (2in. x 2in.) using a Dell C3760dn color printer and scanned with a Fujitsu fi-7280 scanner @ 300 PPI. The live probe images are border crossing photographs collected with a webcam of travelers entering the United States. The border crossing photos often have pose angle and illumination variations.

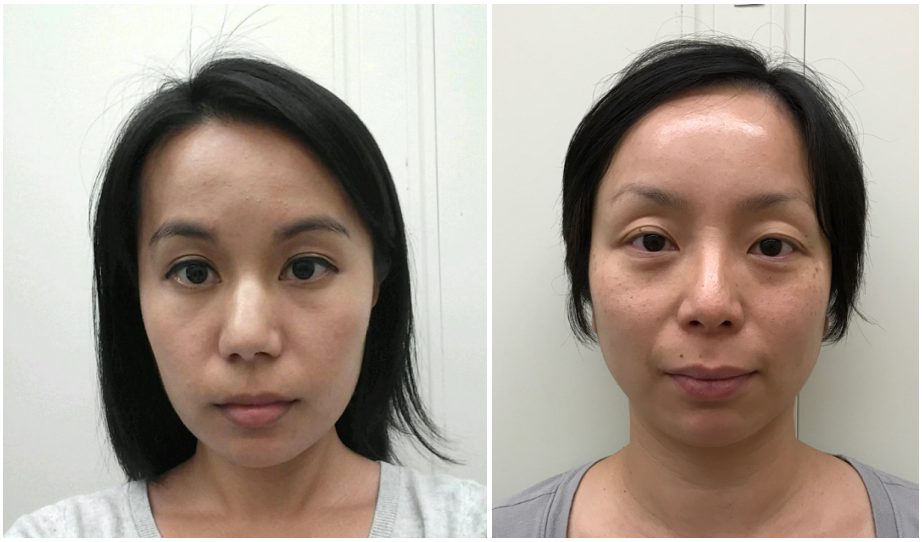
Table 6: Tier 3 datasets: morphs created using manual methods with commercial tools. All morphs are created with two subjects and subject alpha, where known, is 0.5 (i.e., each subject contributed equally to the morph). The image label represents the label that was provided to the algorithm while processing images from the particular dataset.

2.3.4 Other Datasets

Dataset	# Source Images	Image Size	Notes
Mugshots	1047389	499x588, 768x960, 800x1000, 1000x1330	The probe images used to evaluate differential MAD on this dataset are similarly, mugshot-style photos.
Visa	871984	320x320	The visa-like frontal images have geometry in good conformance with the ISO/IEC 19794-5 Full Frontal image type. Pose is generally excellent. The mean interocular distance (IOD) is 61 pixels. All of the images are live capture. The probe images used to evaluate differential MAD on this dataset are webcam photos collected with variations in pose, illumination, and background. See Border crossing webcam probes dataset for additional information.

Border crossing webcam probes	871984	Mostly 340x220	These webcam images are taken with a camera oriented by an attendant toward a cooperating subject. This is done under time constraints, so there are role, pitch and yaw angle variation. The background is not uniform and may contain furniture and windows. There is sometimes perspective distortion due to close range images. The mean IOD is 38 pixels. All of the images are live capture.

Table 7: Other datasets: additional bona fide images used to evaluate morph false detection rate.

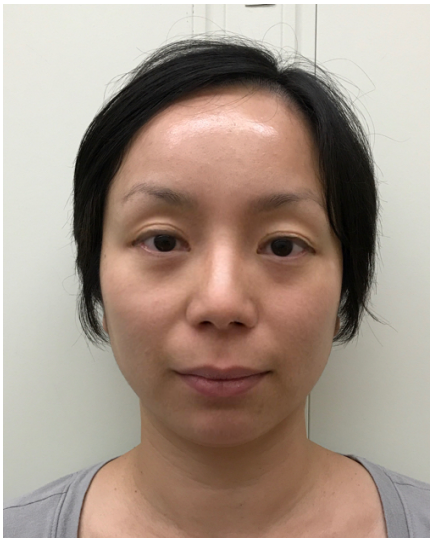


(a) Subject A *(b) Subject B*

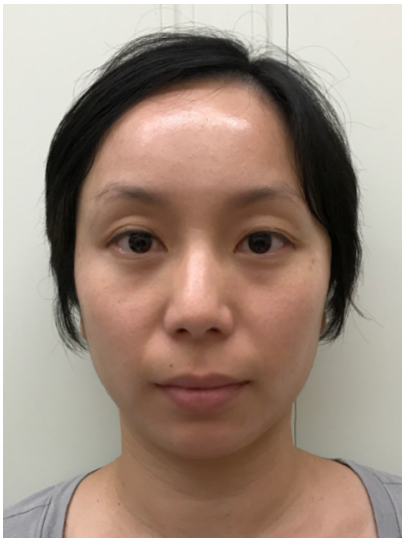
(c) Website *(d) Global* *(e) Local*

(f) Local Morph Colorized Average *(g) Local Morph Colorized Match* *(h) UNIBO Automatic Morphed Face Generation Tool v1.0*

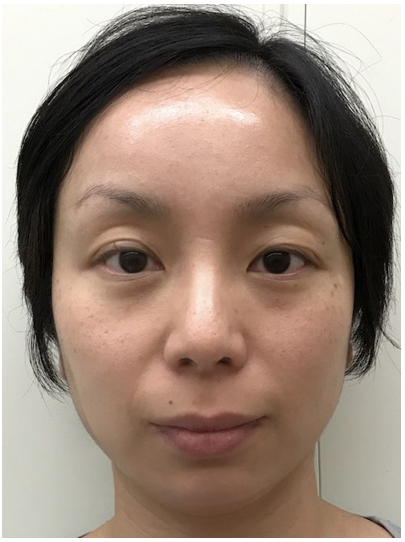
Figure 2: Samples of morphed imagery used in this report.



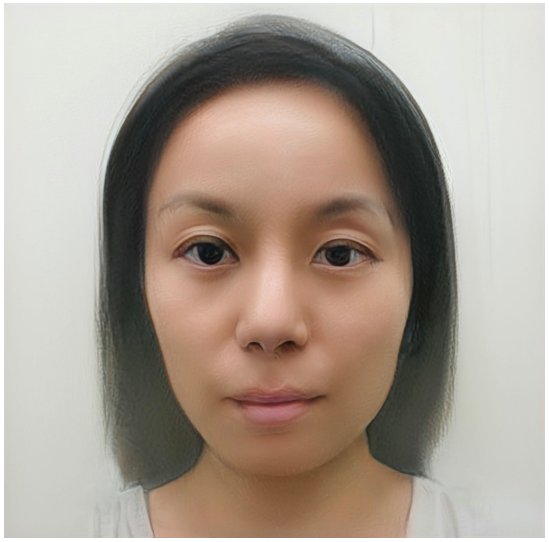
(i) DST



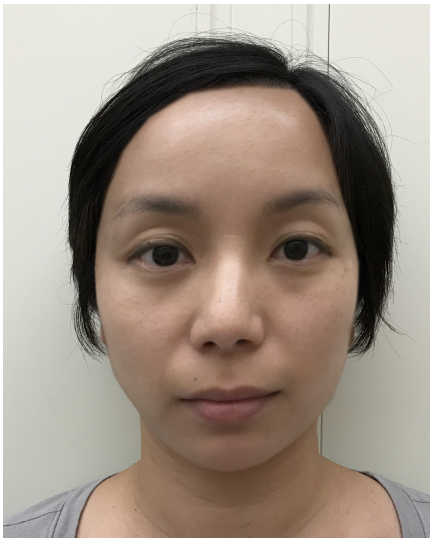
(j) UNIBO Automatic Morphed Face Generation Tool v2.0



(k) Twente

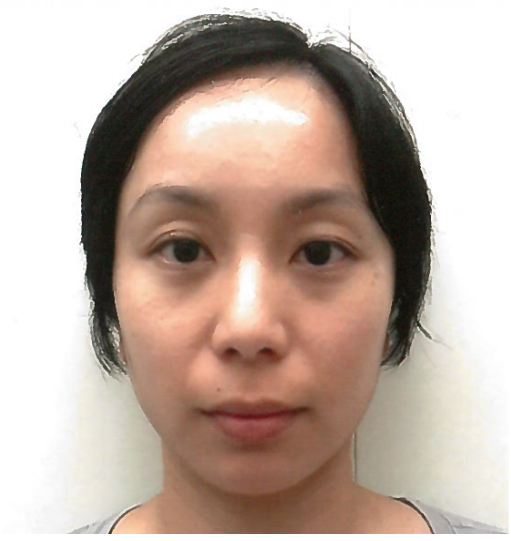


(l) MIPGAN-II



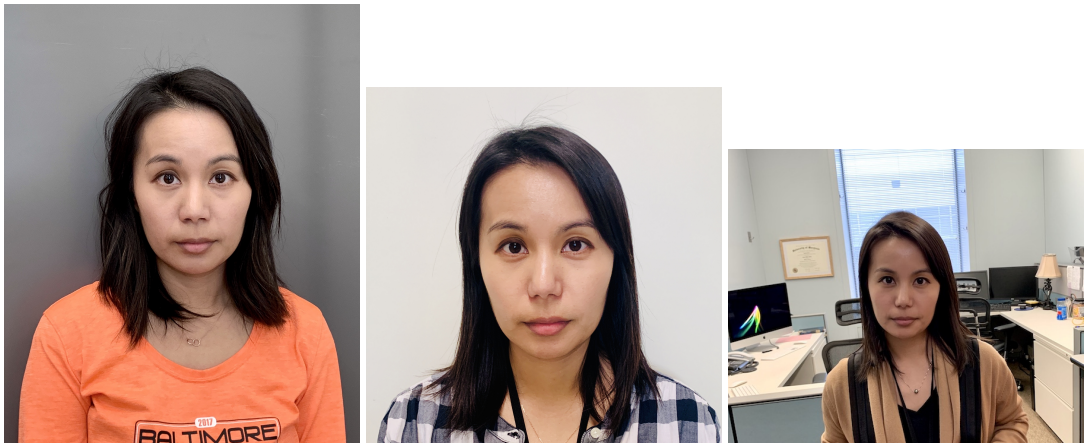
(m) Manual

(n) Lincoln



(o) Print and Scanned

Figure 2: Samples of morphed imagery used in this report. Both subjects of the morphs are NIST employees.



(a) Mugshot *(b) Visa* *(c) Webcam probe*

Figure 3: Samples of bona fide imagery used in this report. The subject in the photos is a NIST employee.

3 Metrics

In this section, we adopt terminology from the presentation attack detection testing standard [33] to quantify morph classification accuracy. Morph detection or attack presentation classification requires submitted algorithms to determine whether a particular image is a morph or not. Given an image, algorithms reported a 1) binary decision on whether the image is a morph or not and 2) a confidence score on $[0, 1]$ representing the algorithm's certainty about whether the image is a morph.

3.1 Morphing Attack Classification Error Rate (MACER)

Using the algorithm's binary decision, MACER is defined as the proportion of morphing attack samples incorrectly classified as bona fide (nonmorph) presentation. This is measured as the number of incorrectly classified morphed images, M, divided by the total number of morphed images, N_m. In the case of algorithm failure to process an image (i.e., the software returns a non-successful return code), those failures are treated as detection of a morphed image with a confidence score of 1 and are incorporated in the calculation of MACER. Additionally, the percentage of morphs that the algorithm "failed to process" is documented as a standalone quantity in this report.

$$\text{MACER} = \frac{M}{N_m} \tag{1}$$

Note that the algorithm's binary decision is based off of some developer-defined internal threshold.

3.2 Bona Fide Sample Classification Error Rate (BSCER)

Similarly, BSCER is defined as the proportion of bona fide (nonmorph) samples incorrectly classified as morphed samples. This is measured as the number of incorrectly classified bona fide images, B, divided by the total number of bona fide images, N_b. In the case of algorithm failure to process an image (i.e., the software returns a non-successful return code), those failures are treated as detection of a morphed image with a confidence score of 1 and are incorporated in the calculation of BSCER. Additionally, the percentage of bona fides that the algorithm "failed to process" are documented as a standalone quantity in this report.

$$\text{BSCER} = \frac{B}{N_b} \tag{2}$$

3.3 Detection Error Tradeoff (DET)

We assess detection accuracy by analyzing the confidence score returned by the algorithm. In this case, the higher the confidence value, the more likely the algorithm thinks it is a morph. A reasonable approach to the detection problem is to classify an image as either a morph or bona fide image by thresholding on its confidence value.

Given N detection scores on bona fide images, b, the BSCER is computed as the proportion above some threshold, T. Similarly, given M detection scores on morphed images, m, the MACER is computed as the proportion below some threshold, T. $H(x)$ is the unit step function [34], and $H(0)$ is taken to be 1.

$$\text{BSCER}(T) = \frac{1}{N}\sum_{i=1}^{N} H(b_i - T), \tag{3}$$

$$\text{MACER}(T) = 1 - \frac{1}{M}\sum_{i=1}^{M} H(m_i - T). \tag{4}$$

In an operational setting, BSCER can be interpreted as the rate of inconvenience for those with a legitimate, bona fide photo on a passport whose photo is being incorrectly detected as a morph. The consequence of such false detections is additional resources required to adjudicate the bona fide photo. Conversely, MACER is the rate that fraud successfully takes place when a morphed photo on a passport is incorrectly classified as a legitimate, bona fide photo (a false negative occurs).

3.3.1 BSCER vs. MACER

Operationally, it is important that morph detection technology produce very low false detection rates given the assumption that most transactions will be on legitimate, bona fide photos. Therefore, the error rate that needs to be controlled is the BSCER, the rate at which bona fide images are falsely classified as morphs. Additional amounts of resources will be required to adjudicate such errors, which drives the need to limit false detections. But given that the technology is still in its infancy and for the purposes of comparing algorithm performance, this document analyzes the trade-off between MACER and BSCER at various thresholds and reports MACER @ BSCER=0.01 and BSCER=0.003.

3.4 Mated Morph Presentation Match Rate (MMPMR)

The mated morph presentation match rate [35] is the fraction of morphed comparisons where all subjects erroneously match the morph above some threshold, T. Formally,

$$\text{MMPMR(T)} = \frac{1}{M}\sum_{m=1}^{M}\left\{\left[\min_{n=1,\ldots,N_m} S_m^n\right] > T\right\} \tag{5}$$

where T is the verification threshold, S_m^n is the mated morph comparison score of the n^{th} subject of morph m, M is the total number of morphed images, and N_m the total number of subjects contributing to morph m.

3.5 False Non-match Rate (FNMR) and False Match Rate (FMR)

For comparisons of bona fide/non-morph photos, given a vector of N genuine scores, u, the false non-match rate is computed as the proportion of scores below some threshold, T:

$$\text{FNMR}(T) = 1 - \frac{1}{N}\sum_{i=1}^{N} H(u_i - T) \tag{6}$$

where $H(x)$ is the unit step function, and $H(0)$ taken to be 1.

Similarly, given a vector of N impostor scores, v, the false match rate is computed as the proportion of scores equal to or above T:

$$\text{FMR}(T) = \frac{1}{N}\sum_{i=1}^{N} H(v_i - T) \tag{7}$$

3.6 Relative Morph Match Rate (RMMR)

The relative morph match rate [35] combines an algorithm's vulnerability to morphs (MMPMR) and its general accuracy on non-morph photos (FNMR), at the same threshold T, into a single measure.

$$\text{RMMR}(T) = 1 + (\text{MMPMR}(T) - (1 - \text{FNMR}(T))) \tag{8}$$

4 Results

4.1 Accuracy Summary

This section provides summary accuracy information of all submitted algorithms against the various datasets that were tested against. Note that for the results in this section, all morphs were created with two subjects only and subject alpha, where known, was 0.5 for each subject (i.e., each subject contributed equally to the morph). Further analysis on morph detection results broken out by subject alpha are in Section 4.10.

4.1.1 BSCER

For each morph dataset, BSCER is evaluated using the methods described below.

- **Single-image morph detection**
  - The first method, BSCER_q, utilizes the source images (where available) that were used to create the morphed images within each dataset. This method attempts to maintain consistent quality between the bona fides and morphs within in each dataset.
  - The second method, BSCER_m, employs the use of a bona fide dataset consisting of approximately 1 million live-capture mugshot photos, which enables the measurement of MACER at low (operationally relevant) BSCER.
  - The third method, BSCER_v, employs the use of a large live-capture bona fide visa dataset composed of approximately 872K images that are in very good conformance with the ISO/IEC 19794-5 Full Frontal image specifications.
- **Two-image differential morph detection**
  - The first method, BSCER_q, utilizes the source images (where available) that were used to create the morphed images within each dataset. The probes are other portrait style images of the subjects.
  - The second method, BSCER_m, employs the use of a bona fide dataset consisting of approximately 1 million live-capture mugshot photos. The probes are other mugshot style images of the subjects. In the future, this method will be augmented to employ the use of webcam-styled probes that better exhibit properties of real-world live-capture probes in operational settings.

– The third method, BSCER_v, employs the use of a large live-capture bona fide visa dataset composed of approximately 872K images that are in very good conformance with the ISO/IEC 19794-5 Full Frontal image specifications. The probes are live-capture webcam photos collected in operational settings with variations in pose, illumination, and background, which more closely mimics, for example, an eGate collection scenario.

4.1.2 Failure to Process

A failure to process occurs when the algorithm software returns a non-successful return code from the morph detection function, indicating that something went wrong while processing the image. Operationally, such failure to process events may trigger secondary processes, which may require additional resources. As such, all occurrences of failure to process by an algorithm are treated as if a morph is detected with the confidence score set to 1 and incorporated into the calculation of both MACER and BSCER. Additionally, failure to process rates are documented independently in the accuracy tables below. For each dataset, Failure to Process (Morphs) is the proportion of morphed photos the software fails on; Failure to Process (Bona Fides)$_q$ is the proportion of source images used as bona fides the software fails to process; and Failure to Process (Bona Fides)$_m$ is the proportion of mugshot photos used as bona fides the software fails to process.

4.1.3 FNMR and FMR

False non-match rates and false match rates are measured from comparisons of live-capture bona fide visa-like portrait enrollment photos with border-crossing webcam probe photos collected in operational settings. 3 225 633 genuine comparisons of 535 329 unique subjects were used to generate FNMR, and 10 000 000 impostor comparisons were used to generate FMR.

4.2 Single-image Morph Detection

A single photo (morph or bona fide) is provided as input to the algorithm.

4.2.1 Tier 1 - Low Quality Morphs

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
idemia-004	Online tool from website	0.005	0.050	0.002	0.005	0.000	0.0000	0.0000	0.003	$0.007^{(1)}$
wvusingle-002	Online tool from website	0.131	0.091	0.138	0.028	0.000	0.0000	0.0000	0.248	$0.427^{(2)}$
visteam-002	Online tool from website	0.330	0.020	0.084	0.015	0.002	0.0000	0.0000	0.380	$0.549^{(3)}$
visteam-004	Online tool from website	0.308	0.011	0.044	0.020	0.002	0.0000	0.0000	0.434	$0.652^{(4)}$
visteam-001	Online tool from website	0.116	0.129	0.465	0.242	0.002	0.0000	0.0000	0.585	$0.739^{(5)}$
visteam-003	Online tool from website	0.434	0.013	0.078	0.026	0.002	0.0000	0.0000	0.587	$0.743^{(6)}$
visteam-000	Online tool from website	0.234	0.093	0.373	0.409	0.002	0.0000	0.0000	0.658	$0.791^{(7)}$
visteamicao-000	Online tool from website	0.786	0.009	0.092	0.013	0.002	0.0000	0.0000	0.817	$0.894^{(8)}$
wvusingle-001	Online tool from website	0.127	0.082	0.113	0.135	0.000	0.0000	0.0000	0.782	$0.921^{(9)}$
visionbox-001	Online tool from website	0.366	0.122	0.289	0.013	0.001	0.0000	0.0003	0.686	$0.927^{(10)}$
visionbox-000	Online tool from website	0.393	0.097	0.253	0.023	0.001	0.0000	0.0003	0.891	$0.946^{(11)}$
hhi-002	Online tool from website	0.368	0.142	0.934	0.079	0.000	0.0000	0.0000	0.874	$0.960^{(12)}$
ntnussl-002	Online tool from website	0.998	0.004	-	0.003	0.002	-	0.0027	0.996	$0.998^{(13)}$

* MACER: This is the rate that morphs are not detected (at some developer-defined threshold). Lower values are better.
** BSCER: This is the rate that bona fides were mistaken for morphs (at some developer-defined threshold). Lower values are better.
For each dataset, the entries are ordered by the metric in the last table column.
Entries with - means results are missing either due to the algorithm not being able to process the entire dataset OR results are still currently being generated.

idemia-003	Online tool from website	0.090	0.047	0.007	0.012	0.000	0.0000	0.0000	0.094	1.000[14]
neurotechnology-000	Online tool from website	0.016	0.844	0.037	0.037	0.000	0.0023	0.0042	0.277	1.000[15]
hhi-001	Online tool from website	0.363	0.149	0.934	0.086	0.004	0.0003	0.0074	0.965	1.000[16]
hdabsif-004	Online tool from website	0.038	0.977	0.001	0.711	0.004	0.0003	0.0082	0.996	1.000[17]
hdafvdet-001	Online tool from website	0.964	0.039	0.039	0.057	0.004	0.0003	0.0089	0.996	1.000[18]
hdalaplace-001	Online tool from website	0.839	0.376	0.003	0.177	0.004	0.0003	0.0073	0.996	1.000[19]
hdaprnu-004	Online tool from website	0.940	0.333	0.695	0.309	0.004	0.0003	0.0073	0.996	1.000[20]
unibo-000	Online tool from website	0.984	0.077	0.006	0.093	0.004	0.0001	0.0045	0.996	1.000[21]
hdafusion-001	Online tool from website	0.785	0.063	0.262	0.137	0.004	0.0004	0.0124	1.000	1.000[22]
hdalbp-005	Online tool from website	0.797	0.174	0.570	0.317	0.007	0.0235	0.1683	1.000	1.000[23]
hdalbp-006	Online tool from website	0.768	0.425	0.786	0.420	0.004	0.0004	0.0110	1.000	1.000[24]
hdaprnu-002	Online tool from website	0.096	0.964	0.987	0.919	0.003	0.0371	0.2877	1.000	1.000[25]

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
idemia-004	Global Morph	0.000	0.004	0.002	0.005	0.000	0.0000	0.0000	0.000	0.000[1]
visionbox-000	Global Morph	0.000	0.961	0.253	0.023	0.000	0.0000	0.0003	0.000	0.000[2]
visionbox-001	Global Morph	0.000	0.961	0.289	0.013	0.000	0.0000	0.0003	0.000	0.000[3]
hhi-002	Global Morph	0.019	0.130	0.934	0.079	0.000	0.0000	0.0000	0.226	0.424[4]
wvusingle-001	Global Morph	0.015	0.118	0.113	0.135	0.000	0.0000	0.0000	0.241	0.435[5]
visteam-002	Global Morph	0.301	0.039	0.084	0.015	0.000	0.0000	0.0000	0.341	0.484[6]
visteam-004	Global Morph	0.217	0.024	0.044	0.020	0.000	0.0000	0.0000	0.317	0.492[7]

visteam-003	Global Morph	0.206	0.075	0.078	0.026	0.000	0.0000	0.0000	0.339	0.507[8]
visteam-001	Global Morph	0.022	0.366	0.465	0.242	0.000	0.0000	0.0000	0.367	0.563[9]
wvusingle-002	Global Morph	0.175	0.008	0.138	0.028	0.000	0.0000	0.0000	0.366	0.611[10]
visteam-000	Global Morph	0.078	0.445	0.373	0.409	0.000	0.0000	0.0000	0.461	0.632[11]
visteamicao-000	Global Morph	0.866	0.000	0.092	0.013	0.000	0.0000	0.0000	0.894	0.956[12]
idemia-003	Global Morph	0.003	0.000	0.007	0.012	0.000	0.0000	0.0000	0.005	1.000[13]
neurotechnology-000	Global Morph	0.126	0.024	0.037	0.037	0.000	0.0023	0.0042	0.218	1.000[14]
hhi-001	Global Morph	0.019	0.130	0.934	0.086	0.000	0.0003	0.0074	0.460	1.000[15]
ntnussl-002	Global Morph	1.000	0.000	-	0.003	0.000	-	0.0027	0.999	1.000[16]
unibo-000	Global Morph	0.798	0.012	0.006	0.093	0.000	0.0001	0.0045	0.999	1.000[17]
hdabsif-004	Global Morph	0.048	0.839	0.001	0.711	0.000	0.0003	0.0082	1.000	1.000[18]
hdafusion-001	Global Morph	0.472	0.051	0.262	0.137	0.000	0.0004	0.0124	1.000	1.000[19]
hdafvdet-001	Global Morph	0.894	0.008	0.039	0.057	0.000	0.0003	0.0089	1.000	1.000[20]
hdalaplace-001	Global Morph	0.876	0.028	0.003	0.177	0.000	0.0003	0.0073	1.000	1.000[21]
hdalbp-005	Global Morph	0.186	0.378	0.570	0.317	0.114	0.0235	0.1683	1.000	1.000[22]
hdalbp-006	Global Morph	0.462	0.106	0.786	0.420	0.000	0.0004	0.0110	1.000	1.000[23]
hdaprnu-002	Global Morph	0.180	0.528	0.987	0.919	0.030	0.0371	0.2877	1.000	1.000[24]
hdaprnu-004	Global Morph	0.984	0.039	0.695	0.309	0.000	0.0003	0.0073	1.000	1.000[25]

4.2.2 Tier 2 - Automated Morphs

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
idemia-004	Local Morph	0.000	0.004	0.002	0.005	0.000	0.0000	0.0000	0.000	0.000[1]
visionbox-000	Local Morph	0.000	0.961	0.253	0.023	0.000	0.0000	0.0003	0.000	0.000[2]
visionbox-001	Local Morph	0.000	0.961	0.289	0.013	0.000	0.0000	0.0003	0.000	0.000[3]
visteam-004	Local Morph	0.110	0.024	0.044	0.020	0.000	0.0000	0.0000	0.176	0.354[4]
visteam-003	Local Morph	0.124	0.075	0.078	0.026	0.000	0.0000	0.0000	0.232	0.397[5]

visteam-002	Local Morph	0.211	0.039	0.084	0.015	0.000	0.0000	0.0000	0.261	0.423[6]
hhi-002	Local Morph	0.023	0.130	0.934	0.079	0.000	0.0000	0.0000	0.239	0.446[7]
visteam-000	Local Morph	0.029	0.445	0.373	0.409	0.000	0.0000	0.0000	0.304	0.457[8]
visteam-001	Local Morph	0.012	0.366	0.465	0.242	0.000	0.0000	0.0000	0.308	0.517[9]
wvusingle-001	Local Morph	0.042	0.118	0.113	0.135	0.000	0.0000	0.0000	0.407	0.638[10]
wvusingle-002	Local Morph	0.421	0.008	0.138	0.028	0.000	0.0000	0.0000	0.644	0.815[11]
visteamicao-000	Local Morph	0.716	0.000	0.092	0.013	0.000	0.0000	0.0000	0.745	0.893[12]
idemia-003	Local Morph	0.008	0.000	0.007	0.012	0.000	0.0000	0.0000	0.009	1.000[13]
neurotechnology-000	Local Morph	0.067	0.024	0.037	0.037	0.000	0.0023	0.0042	0.133	1.000[14]
hhi-001	Local Morph	0.023	0.130	0.934	0.086	0.000	0.0003	0.0074	0.475	1.000[15]
ntnussl-002	Local Morph	1.000	0.000	-	0.003	0.000	-	0.0027	0.999	1.000[16]
unibo-000	Local Morph	0.839	0.012	0.006	0.093	0.000	0.0001	0.0045	0.999	1.000[17]
hdabsif-004	Local Morph	0.112	0.839	0.001	0.711	0.000	0.0003	0.0082	1.000	1.000[18]
hdafusion-001	Local Morph	0.413	0.051	0.262	0.137	0.001	0.0004	0.0124	1.000	1.000[19]
hdafvdet-001	Local Morph	0.907	0.008	0.039	0.057	0.000	0.0003	0.0089	1.000	1.000[20]
hdalaplace-001	Local Morph	0.891	0.028	0.003	0.177	0.000	0.0003	0.0073	1.000	1.000[21]
hdalbp-005	Local Morph	0.258	0.378	0.570	0.317	0.055	0.0235	0.1683	1.000	1.000[22]
hdalbp-006	Local Morph	0.431	0.106	0.786	0.420	0.001	0.0004	0.0110	1.000	1.000[23]
hdaprnu-002	Local Morph	0.138	0.528	0.987	0.919	0.063	0.0371	0.2877	1.000	1.000[24]
hdaprnu-004	Local Morph	0.984	0.039	0.695	0.309	0.000	0.0003	0.0073	1.000	1.000[25]

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
idemia-004	Local Morph Colorized Average	0.000	0.004	0.002	0.005	0.000	0.0000	0.0000	0.000	0.000[1]
visionbox-000	Local Morph Colorized Average	0.000	0.961	0.253	0.023	0.000	0.0000	0.0003	0.000	0.000[2]

visionbox-001	Local Morph Colorized Average	0.000	0.961	0.289	0.013	0.000	0.0000	0.0003	0.000	$0.000^{(3)}$
hhi-002	Local Morph Colorized Average	0.022	0.130	0.934	0.079	0.000	0.0000	0.0000	0.236	$0.434^{(4)}$
visteam-000	Local Morph Colorized Average	0.054	0.445	0.373	0.409	0.000	0.0000	0.0000	0.380	$0.540^{(5)}$
visteam-002	Local Morph Colorized Average	0.311	0.039	0.084	0.015	0.000	0.0000	0.0000	0.354	$0.545^{(6)}$
wvusingle-001	Local Morph Colorized Average	0.025	0.118	0.113	0.135	0.000	0.0000	0.0000	0.334	$0.570^{(7)}$
visteam-004	Local Morph Colorized Average	0.289	0.024	0.044	0.020	0.000	0.0000	0.0000	0.384	$0.576^{(8)}$
visteam-001	Local Morph Colorized Average	0.008	0.366	0.465	0.242	0.000	0.0000	0.0000	0.348	$0.590^{(9)}$
visteam-003	Local Morph Colorized Average	0.273	0.075	0.078	0.026	0.000	0.0000	0.0000	0.421	$0.602^{(10)}$
wvusingle-002	Local Morph Colorized Average	0.355	0.008	0.138	0.028	0.000	0.0000	0.0000	0.590	$0.808^{(11)}$
visteamicao-000	Local Morph Colorized Average	0.769	0.000	0.092	0.013	0.000	0.0000	0.0000	0.805	$0.921^{(12)}$
idemia-003	Local Morph Colorized Average	0.029	0.000	0.007	0.012	0.000	0.0000	0.0000	0.032	$1.000^{(13)}$
neurotechnology-000	Local Morph Colorized Average	0.305	0.024	0.037	0.037	0.000	0.0023	0.0042	0.410	$1.000^{(14)}$
hhi-001	Local Morph Colorized Average	0.022	0.130	0.934	0.086	0.000	0.0003	0.0074	0.475	$1.000^{(15)}$

unibo-000	Local Morph Colorized Average	0.836	0.012	0.006	0.093	0.000	0.0001	0.0045	0.999	1.000[16]
hdabsif-004	Local Morph Colorized Average	0.118	0.839	0.001	0.711	0.000	0.0003	0.0082	1.000	1.000[17]
hdafusion-001	Local Morph Colorized Average	0.413	0.051	0.262	0.137	0.001	0.0004	0.0124	1.000	1.000[18]
hdafvdet-001	Local Morph Colorized Average	0.935	0.008	0.039	0.057	0.000	0.0003	0.0089	1.000	1.000[19]
hdalaplace-001	Local Morph Colorized Average	0.887	0.028	0.003	0.177	0.000	0.0003	0.0073	1.000	1.000[20]
hdalbp-005	Local Morph Colorized Average	0.238	0.378	0.570	0.317	0.084	0.0235	0.1683	1.000	1.000[21]
hdalbp-006	Local Morph Colorized Average	0.432	0.106	0.786	0.420	0.000	0.0004	0.0110	1.000	1.000[22]
hdaprnu-002	Local Morph Colorized Average	0.121	0.528	0.987	0.919	0.042	0.0371	0.2877	1.000	1.000[23]
hdaprnu-004	Local Morph Colorized Average	0.981	0.039	0.695	0.309	0.000	0.0003	0.0073	1.000	1.000[24]
ntnussl-002	Local Morph Colorized Average	1.000	0.000	-	0.003	0.000	-	0.0027	1.000	1.000[25]

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
idemia-004	Local Morph Colorized Match	0.000	0.004	0.002	0.005	0.000	0.0000	0.0000	0.000	0.000[1]
visionbox-000	Local Morph Colorized Match	0.000	0.961	0.253	0.023	0.000	0.0000	0.0003	0.000	0.000[2]
visionbox-001	Local Morph Colorized Match	0.000	0.961	0.289	0.013	0.000	0.0000	0.0003	0.000	0.000[3]

visteam-000	Local Morph Colorized Match	0.078	0.445	0.373	0.409	0.000	0.0000	0.0000	0.447	$0.617^{(4)}$
visteam-004	Local Morph Colorized Match	0.374	0.024	0.044	0.020	0.000	0.0000	0.0000	0.466	$0.643^{(5)}$
visteam-003	Local Morph Colorized Match	0.342	0.075	0.078	0.026	0.000	0.0000	0.0000	0.490	$0.674^{(6)}$
hhi-002	Local Morph Colorized Match	0.187	0.130	0.934	0.079	0.000	0.0000	0.0000	0.574	$0.744^{(7)}$
visteam-002	Local Morph Colorized Match	0.559	0.039	0.084	0.015	0.000	0.0000	0.0000	0.614	$0.762^{(8)}$
visteam-001	Local Morph Colorized Match	0.091	0.366	0.465	0.242	0.000	0.0000	0.0000	0.620	$0.788^{(9)}$
wvusingle-001	Local Morph Colorized Match	0.236	0.118	0.113	0.135	0.000	0.0000	0.0000	0.724	$0.862^{(10)}$
wvusingle-002	Local Morph Colorized Match	0.790	0.008	0.138	0.028	0.000	0.0000	0.0000	0.897	$0.959^{(11)}$
visteamicao-000	Local Morph Colorized Match	0.880	0.000	0.092	0.013	0.000	0.0000	0.0000	0.903	$0.963^{(12)}$
idemia-003	Local Morph Colorized Match	0.053	0.000	0.007	0.012	0.000	0.0000	0.0000	0.060	$1.000^{(13)}$
neurotechnology-000	Local Morph Colorized Match	0.495	0.024	0.037	0.037	0.000	0.0023	0.0042	0.624	$1.000^{(14)}$
hhi-001	Local Morph Colorized Match	0.187	0.130	0.934	0.086	0.000	0.0003	0.0074	0.779	$1.000^{(15)}$
hdabsif-004	Local Morph Colorized Match	0.105	0.839	0.001	0.711	0.000	0.0003	0.0082	1.000	$1.000^{(16)}$

hdafusion-001	Local Morph Colorized Match	0.489	0.051	0.262	0.137	0.001	0.0004	0.0124	1.000	1.000[17]
hdafvdet-001	Local Morph Colorized Match	0.930	0.008	0.039	0.057	0.000	0.0003	0.0089	1.000	1.000[18]
hdalaplace-001	Local Morph Colorized Match	0.928	0.028	0.003	0.177	0.000	0.0003	0.0073	1.000	1.000[19]
hdalbp-005	Local Morph Colorized Match	0.296	0.378	0.570	0.317	0.063	0.0235	0.1683	1.000	1.000[20]
hdalbp-006	Local Morph Colorized Match	0.535	0.106	0.786	0.420	0.000	0.0004	0.0110	1.000	1.000[21]
hdaprnu-002	Local Morph Colorized Match	0.285	0.528	0.987	0.919	0.051	0.0371	0.2877	1.000	1.000[22]
hdaprnu-004	Local Morph Colorized Match	0.985	0.039	0.695	0.309	0.000	0.0003	0.0073	1.000	1.000[23]
ntnussl-002	Local Morph Colorized Match	1.000	0.000	-	0.003	0.000	-	0.0027	1.000	1.000[24]
unibo-000	Local Morph Colorized Match	0.947	0.012	0.006	0.093	0.000	0.0001	0.0045	1.000	1.000[25]

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
hhi-002	UNIBO Automatic Morphed Face Generation Tool v1.0	0.000	0.734	0.934	0.079	0.000	0.0000	0.0000	0.000	0.000[1]
idemia-004	UNIBO Automatic Morphed Face Generation Tool v1.0	0.000	0.000	0.002	0.005	0.000	0.0000	0.0000	0.000	0.000[2]

visionbox-000	UNIBO Automatic Morphed Face Generation Tool v1.0	0.000	0.250	0.253	0.023	0.000	0.0000	0.0003	0.000	$0.000^{(3)}$
visionbox-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.000	0.078	0.289	0.013	0.000	0.0000	0.0003	0.000	$0.000^{(4)}$
wvusingle-002	UNIBO Automatic Morphed Face Generation Tool v1.0	0.018	0.125	0.138	0.028	0.000	0.0000	0.0000	0.075	$0.179^{(5)}$
visteamicao-000	UNIBO Automatic Morphed Face Generation Tool v1.0	0.082	0.016	0.092	0.013	0.000	0.0000	0.0000	0.095	$0.212^{(6)}$
visteam-004	UNIBO Automatic Morphed Face Generation Tool v1.0	0.130	0.016	0.044	0.020	0.000	0.0000	0.0000	0.175	$0.270^{(7)}$
visteam-000	UNIBO Automatic Morphed Face Generation Tool v1.0	0.028	0.469	0.373	0.409	0.000	0.0000	0.0000	0.253	$0.382^{(8)}$
visteam-003	UNIBO Automatic Morphed Face Generation Tool v1.0	0.221	0.062	0.078	0.026	0.000	0.0000	0.0000	0.356	$0.517^{(9)}$
visteam-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.006	0.438	0.465	0.242	0.000	0.0000	0.0000	0.349	$0.538^{(10)}$
visteam-002	UNIBO Automatic Morphed Face Generation Tool v1.0	0.364	0.047	0.084	0.015	0.000	0.0000	0.0000	0.407	$0.548^{(11)}$

wvusingle-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.075	0.219	0.113	0.135	0.000	0.0000	0.0000	0.406	0.622[12]
hhi-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.000	0.734	0.934	0.086	0.000	0.0003	0.0074	0.000	1.000[13]
idemia-003	UNIBO Automatic Morphed Face Generation Tool v1.0	0.000	0.016	0.007	0.012	0.000	0.0000	0.0000	0.001	1.000[14]
neurotechnology-000	UNIBO Automatic Morphed Face Generation Tool v1.0	0.022	0.031	0.037	0.037	0.000	0.0023	0.0042	0.043	1.000[15]
unibo-000	UNIBO Automatic Morphed Face Generation Tool v1.0	0.000	0.641	0.006	0.093	0.000	0.0001	0.0045	0.087	1.000[16]
hdafvdet-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.090	0.109	0.039	0.057	0.000	0.0003	0.0089	0.994	1.000[17]
ntnussl-002	UNIBO Automatic Morphed Face Generation Tool v1.0	1.000	0.000	-	0.003	0.000	-	0.0027	0.999	1.000[18]
hdabsif-004	UNIBO Automatic Morphed Face Generation Tool v1.0	0.000	1.000	0.001	0.711	0.000	0.0003	0.0082	1.000	1.000[19]
hdafusion-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.043	0.578	0.262	0.137	0.000	0.0004	0.0124	1.000	1.000[20]

hdalaplace-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.447	0.031	0.003	0.177	0.000	0.0003	0.0073	1.000	$1.000^{(21)}$
hdalbp-005	UNIBO Automatic Morphed Face Generation Tool v1.0	0.146	0.500	0.570	0.317	0.075	0.0235	0.1683	1.000	$1.000^{(22)}$
hdalbp-006	UNIBO Automatic Morphed Face Generation Tool v1.0	0.019	0.469	0.786	0.420	0.000	0.0004	0.0110	1.000	$1.000^{(23)}$
hdaprnu-002	UNIBO Automatic Morphed Face Generation Tool v1.0	0.000	0.906	0.987	0.919	0.000	0.0371	0.2877	1.000	$1.000^{(24)}$
hdaprnu-004	UNIBO Automatic Morphed Face Generation Tool v1.0	0.510	0.047	0.695	0.309	0.000	0.0003	0.0073	1.000	$1.000^{(25)}$

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
visionbox-000	Visa-Border	0.000	0.178	0.253	0.023	0.0000	0.0000	0.0003	0.000	$0.000^{(1)}$
visionbox-001	Visa-Border	0.000	0.209	0.289	0.013	0.0000	0.0000	0.0003	0.000	$0.000^{(2)}$
idemia-004	Visa-Border	0.001	0.002	0.002	0.005	0.0000	0.0000	0.0000	0.000	$0.001^{(3)}$
hhi-002	Visa-Border	0.000	0.933	0.934	0.079	0.0000	0.0000	0.0000	0.007	$0.024^{(4)}$
visteamicao-000	Visa-Border	0.261	0.082	0.092	0.013	0.0000	0.0000	0.0000	0.291	$0.450^{(5)}$
visteam-002	Visa-Border	0.365	0.062	0.084	0.015	0.0000	0.0000	0.0000	0.414	$0.565^{(6)}$
visteam-001	Visa-Border	0.077	0.353	0.465	0.242	0.0000	0.0000	0.0000	0.518	$0.667^{(7)}$
visteam-004	Visa-Border	0.392	0.039	0.044	0.020	0.0000	0.0000	0.0000	0.507	$0.683^{(8)}$
visteam-003	Visa-Border	0.420	0.052	0.078	0.026	0.0000	0.0000	0.0000	0.555	$0.707^{(9)}$
wvusingle-002	Visa-Border	0.253	0.117	0.138	0.028	0.0000	0.0000	0.0000	0.542	$0.810^{(10)}$

visteam-000	Visa-Border	0.262	0.308	0.373	0.409	0.0000	0.0000	0.0000	0.686	0.811[11]
hhi-001	Visa-Border	0.000	0.933	0.934	0.086	0.0000	0.0003	0.0074	0.029	1.000[12]
idemia-003	Visa-Border	0.026	0.006	0.007	0.012	0.0000	0.0000	0.0000	0.030	1.000[13]
neurotechnology-000	Visa-Border	0.258	0.030	0.037	0.037	0.0049	0.0023	0.0042	0.353	1.000[14]
hdaprnu-004	Visa-Border	0.009	0.802	0.695	0.309	0.0000	0.0003	0.0073	0.823	1.000[15]
ntnussl-002	Visa-Border	1.000	-	-	0.003	0.0000	-	0.0027	0.990	1.000[16]
wvusingle-001	Visa-Border	0.947	0.076	0.113	0.135	0.0000	0.0000	0.0000	0.998	1.000[17]
unibo-000	Visa-Border	0.536	0.013	0.006	0.093	0.0000	0.0001	0.0045	0.999	1.000[18]
hdabsif-004	Visa-Border	1.000	0.000	0.001	0.711	0.0000	0.0003	0.0082	1.000	1.000[19]
hdafusion-001	Visa-Border	0.340	0.167	0.262	0.137	0.0000	0.0004	0.0124	1.000	1.000[20]
hdafvdet-001	Visa-Border	0.853	0.034	0.039	0.057	0.0000	0.0003	0.0089	1.000	1.000[21]
hdalaplace-001	Visa-Border	0.938	0.004	0.003	0.177	0.0000	0.0003	0.0073	1.000	1.000[22]
hdalbp-005	Visa-Border	0.041	0.600	0.570	0.317	0.0304	0.0235	0.1683	1.000	1.000[23]
hdalbp-006	Visa-Border	0.004	0.847	0.786	0.420	0.0000	0.0004	0.0110	1.000	1.000[24]
hdaprnu-002	Visa-Border	0.000	0.992	0.987	0.919	0.0003	0.0371	0.2877	1.000	1.000[25]

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
idemia-004	UNIBO Automatic Morphed Face Generation Tool v2.0	0.000	0.000	0.002	0.005	0.000	0.0000	0.0000	0.000	0.000[1]
visionbox-000	UNIBO Automatic Morphed Face Generation Tool v2.0	0.000	0.250	0.253	0.023	0.000	0.0000	0.0003	0.000	0.000[2]
visionbox-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.000	0.078	0.289	0.013	0.000	0.0000	0.0003	0.000	0.000[3]

hhi-002	UNIBO Automatic Morphed Face Generation Tool v2.0	0.000	0.734	0.934	0.079	0.000	0.0000	0.0000	0.002	$0.009^{(4)}$
visteamicao-000	UNIBO Automatic Morphed Face Generation Tool v2.0	0.156	0.016	0.092	0.013	0.000	0.0000	0.0000	0.192	$0.358^{(5)}$
wvusingle-002	UNIBO Automatic Morphed Face Generation Tool v2.0	0.134	0.125	0.138	0.028	0.000	0.0000	0.0000	0.229	$0.381^{(6)}$
visteam-004	UNIBO Automatic Morphed Face Generation Tool v2.0	0.218	0.016	0.044	0.020	0.000	0.0000	0.0000	0.287	$0.410^{(7)}$
visteam-000	UNIBO Automatic Morphed Face Generation Tool v2.0	0.052	0.469	0.373	0.409	0.000	0.0000	0.0000	0.376	$0.548^{(8)}$
wvusingle-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.156	0.219	0.113	0.135	0.000	0.0000	0.0000	0.494	$0.679^{(9)}$
visteam-003	UNIBO Automatic Morphed Face Generation Tool v2.0	0.359	0.062	0.078	0.026	0.000	0.0000	0.0000	0.517	$0.692^{(10)}$
visteam-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.034	0.438	0.465	0.242	0.000	0.0000	0.0000	0.555	$0.736^{(11)}$
visteam-002	UNIBO Automatic Morphed Face Generation Tool v2.0	0.544	0.047	0.084	0.015	0.000	0.0000	0.0000	0.595	$0.748^{(12)}$

idemia-003	UNIBO Automatic Morphed Face Generation Tool v2.0	0.006	0.016	0.007	0.012	0.000	0.0000	0.0000	0.007	1.000[13]
hhi-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.000	0.734	0.934	0.086	0.000	0.0003	0.0074	0.011	1.000[14]
unibo-000	UNIBO Automatic Morphed Face Generation Tool v2.0	0.002	0.641	0.006	0.093	0.000	0.0001	0.0045	0.156	1.000[15]
neurotechnology-000	UNIBO Automatic Morphed Face Generation Tool v2.0	0.164	0.031	0.037	0.037	0.000	0.0023	0.0042	0.228	1.000[16]
hdafvdet-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.132	0.109	0.039	0.057	0.000	0.0003	0.0089	0.993	1.000[17]
hdabsif-004	UNIBO Automatic Morphed Face Generation Tool v2.0	0.000	1.000	0.001	0.711	0.000	0.0003	0.0082	1.000	1.000[18]
hdafusion-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.050	0.578	0.262	0.137	0.000	0.0004	0.0124	1.000	1.000[19]
hdalaplace-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.384	0.031	0.003	0.177	0.000	0.0003	0.0073	1.000	1.000[20]
hdalbp-005	UNIBO Automatic Morphed Face Generation Tool v2.0	0.194	0.500	0.570	0.317	0.069	0.0235	0.1683	1.000	1.000[21]

hdalbp-006	UNIBO Automatic Morphed Face Generation Tool v2.0	0.030	0.469	0.786	0.420	0.000	0.0004	0.0110	1.000	$1.000^{(22)}$
hdaprnu-002	UNIBO Automatic Morphed Face Generation Tool v2.0	0.000	0.906	0.987	0.919	0.000	0.0371	0.2877	1.000	$1.000^{(23)}$
hdaprnu-004	UNIBO Automatic Morphed Face Generation Tool v2.0	0.389	0.047	0.695	0.309	0.000	0.0003	0.0073	1.000	$1.000^{(24)}$
ntnussl-002	UNIBO Automatic Morphed Face Generation Tool v2.0	1.000	0.000	-	0.003	0.000	-	0.0027	1.000	$1.000^{(25)}$

Algorithm	Dataset	MACER*	$BSCER_q$ **	$BSCER_v$ ** (visa)	$BSCER_m$ ** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
idemia-004	Twente	0.001	0.000	0.002	0.005	0.000	0.0000	0.0000	0.001	$0.001^{(1)}$
visionbox-000	Twente	0.001	0.250	0.253	0.023	0.000	0.0000	0.0003	0.001	$0.001^{(2)}$
visionbox-001	Twente	0.001	0.078	0.289	0.013	0.000	0.0000	0.0003	0.001	$0.001^{(3)}$
hhi-002	Twente	0.000	0.734	0.934	0.079	0.000	0.0000	0.0000	0.002	$0.002^{(4)}$
wvusingle-002	Twente	0.017	0.125	0.138	0.028	0.000	0.0000	0.0000	0.060	$0.158^{(5)}$
visteamicao-000	Twente	0.112	0.016	0.092	0.013	0.000	0.0000	0.0000	0.128	$0.263^{(6)}$
visteam-004	Twente	0.149	0.016	0.044	0.020	0.000	0.0000	0.0000	0.204	$0.318^{(7)}$
wvusingle-001	Twente	0.012	0.219	0.113	0.135	0.000	0.0000	0.0000	0.209	$0.385^{(8)}$
visteam-000	Twente	0.028	0.469	0.373	0.409	0.000	0.0000	0.0000	0.265	$0.406^{(9)}$
visteam-003	Twente	0.358	0.062	0.078	0.026	0.000	0.0000	0.0000	0.493	$0.666^{(10)}$
visteam-002	Twente	0.510	0.047	0.084	0.015	0.000	0.0000	0.0000	0.553	$0.677^{(11)}$
visteam-001	Twente	0.042	0.438	0.465	0.242	0.000	0.0000	0.0000	0.534	$0.705^{(12)}$
idemia-003	Twente	0.001	0.016	0.007	0.012	0.000	0.0000	0.0000	0.001	$1.000^{(13)}$
hhi-001	Twente	0.000	0.734	0.934	0.086	0.000	0.0003	0.0074	0.002	$1.000^{(14)}$
neurotechnology-000	Twente	0.058	0.031	0.037	0.037	0.000	0.0023	0.0042	0.101	$1.000^{(15)}$
unibo-000	Twente	0.002	0.641	0.006	0.093	0.000	0.0001	0.0045	0.183	$1.000^{(16)}$
hdafvdet-001	Twente	0.451	0.109	0.039	0.057	0.000	0.0003	0.0089	0.991	$1.000^{(17)}$
ntnussl-002	Twente	1.000	0.000	-	0.003	0.000	-	0.0027	0.998	$1.000^{(18)}$
hdabsif-004	Twente	0.000	1.000	0.001	0.711	0.000	0.0003	0.0082	1.000	$1.000^{(19)}$
hdafusion-001	Twente	0.072	0.578	0.262	0.137	0.000	0.0004	0.0124	1.000	$1.000^{(20)}$
hdalaplace-001	Twente	0.984	0.031	0.003	0.177	0.000	0.0003	0.0073	1.000	$1.000^{(21)}$
hdalbp-005	Twente	0.179	0.500	0.570	0.317	0.266	0.0235	0.1683	1.000	$1.000^{(22)}$
hdalbp-006	Twente	0.115	0.469	0.786	0.420	0.000	0.0004	0.0110	1.000	$1.000^{(23)}$
hdaprnu-002	Twente	0.069	0.906	0.987	0.919	0.002	0.0371	0.2877	1.000	$1.000^{(24)}$

hdaprnu-004	Twente	0.982	0.047	0.695	0.309	0.000	0.0003	0.0073	1.000	1.000[25]

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
hhi-002	MIPGAN-II	0.000	0.734	0.934	0.079	0.000	0.0000	0.0000	0.000	0.000[1]
idemia-004	MIPGAN-II	0.000	0.000	0.002	0.005	0.000	0.0000	0.0000	0.000	0.000[2]
visionbox-000	MIPGAN-II	0.000	0.250	0.253	0.023	0.000	0.0000	0.0003	0.000	0.000[3]
visionbox-001	MIPGAN-II	0.000	0.078	0.289	0.013	0.000	0.0000	0.0003	0.000	0.000[4]
wvusingle-002	MIPGAN-II	0.019	0.125	0.138	0.028	0.000	0.0000	0.0000	0.111	0.298[5]
wvusingle-001	MIPGAN-II	0.008	0.219	0.113	0.135	0.000	0.0000	0.0000	0.200	0.388[6]
hhi-001	MIPGAN-II	0.000	0.734	0.934	0.086	0.000	0.0003	0.0074	0.000	0.734[7]
visteam-000	MIPGAN-II	0.157	0.469	0.373	0.409	0.000	0.0000	0.0000	0.639	0.793[8]
visteam-004	MIPGAN-II	0.745	0.016	0.044	0.020	0.000	0.0000	0.0000	0.819	0.899[9]
visteamicao-000	MIPGAN-II	0.715	0.016	0.092	0.013	0.000	0.0000	0.0000	0.761	0.907[10]
visteam-003	MIPGAN-II	0.835	0.062	0.078	0.026	0.000	0.0000	0.0000	0.902	0.959[11]
visteam-002	MIPGAN-II	0.928	0.047	0.084	0.015	0.000	0.0000	0.0000	0.949	0.978[12]
visteam-001	MIPGAN-II	0.434	0.438	0.465	0.242	0.000	0.0000	0.0000	0.960	0.992[13]
idemia-003	MIPGAN-II	0.000	0.016	0.007	0.012	0.000	0.0000	0.0000	0.000	1.000[14]
neurotechnology-000	MIPGAN-II	0.064	0.031	0.037	0.037	0.001	0.0023	0.0042	0.304	1.000[15]
unibo-000	MIPGAN-II	0.041	0.641	0.006	0.093	0.000	0.0001	0.0045	0.810	1.000[16]
hdafvdet-001	MIPGAN-II	0.802	0.109	0.039	0.057	0.000	0.0003	0.0089	0.996	1.000[17]
ntnussl-002	MIPGAN-II	1.000	0.000	-	0.003	0.000	-	0.0027	0.998	1.000[18]
hdabsif-004	MIPGAN-II	0.000	1.000	0.001	0.711	0.000	0.0003	0.0082	1.000	1.000[19]
hdafusion-001	MIPGAN-II	0.737	0.578	0.262	0.137	0.000	0.0004	0.0124	1.000	1.000[20]
hdalaplace-001	MIPGAN-II	0.960	0.031	0.003	0.177	0.000	0.0003	0.0073	1.000	1.000[21]
hdalbp-005	MIPGAN-II	0.299	0.500	0.570	0.317	0.000	0.0235	0.1683	1.000	1.000[22]
hdalbp-006	MIPGAN-II	0.756	0.469	0.786	0.420	0.000	0.0004	0.0110	1.000	1.000[23]
hdaprnu-002	MIPGAN-II	0.046	0.906	0.987	0.919	0.000	0.0371	0.2877	1.000	1.000[24]

hdaprnu-004	MIPGAN-II	0.989	0.047	0.695	0.309	0.000	0.0003	0.0073	1.000	1.000[25]

4.2.3 Tier 3 - High Quality Morphs

Algorithm	Dataset	MACER*	BSCER$_q$**	BSCER$_v$** (visa)	BSCER$_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process (Bona Fides)$_v$	Failure to Process (Bona Fides)$_m$	MACER @ BSCER$_m$ =0.01	MACER @ BSCER$_m$ =0.003
idemia-004	Manual	0.455	0.012	0.002	0.005	0.000	0.0000	0.0000	0.375	0.526[1]
visteam-003	Manual	0.836	0.015	0.078	0.026	0.000	0.0000	0.0000	0.926	0.966[2]
visteam-004	Manual	0.898	0.004	0.044	0.020	0.000	0.0000	0.0000	0.932	0.966[3]
visteam-002	Manual	0.932	0.012	0.084	0.015	0.000	0.0000	0.0000	0.950	0.978[4]
visteam-000	Manual	0.653	0.252	0.373	0.409	0.000	0.0000	0.0000	0.954	0.981[5]
hhi-002	Manual	0.724	0.259	0.934	0.079	0.000	0.0000	0.0000	0.963	0.981[6]
visteam-001	Manual	0.743	0.142	0.465	0.242	0.000	0.0000	0.0000	0.969	0.985[7]
visteamicao-000	Manual	0.975	0.008	0.092	0.013	0.000	0.0000	0.0000	0.975	0.988[8]
wvusingle-002	Manual	0.950	0.038	0.138	0.028	0.000	0.0000	0.0000	0.975	0.991[9]
wvusingle-001	Manual	0.892	0.036	0.113	0.135	0.000	0.0000	0.0000	0.991	0.991[10]
visionbox-001	Manual	0.975	0.159	0.289	0.013	0.000	0.0000	0.0003	0.991	0.997[11]
visionbox-000	Manual	0.972	0.161	0.253	0.023	0.000	0.0000	0.0003	0.994	0.997[12]
idemia-003	Manual	0.644	0.032	0.007	0.012	0.000	0.0000	0.0000	0.659	1.000[13]
neurotechnology-000	Manual	0.858	0.114	0.037	0.037	0.000	0.0023	0.0042	0.966	1.000[14]
hhi-001	Manual	0.724	0.265	0.934	0.086	0.000	0.0003	0.0074	0.985	1.000[15]
ntnussl-002	Manual	1.000	0.013	-	0.003	0.000	-	0.0027	0.985	1.000[16]
hdabsif-004	Manual	0.195	0.545	0.001	0.711	0.000	0.0003	0.0082	1.000	1.000[17]
hdafusion-001	Manual	0.954	0.051	0.262	0.137	0.000	0.0004	0.0124	1.000	1.000[18]
hdafvdet-001	Manual	0.985	0.017	0.039	0.057	0.000	0.0003	0.0089	1.000	1.000[19]
hdalaplace-001	Manual	0.963	0.067	0.003	0.177	0.000	0.0003	0.0073	1.000	1.000[20]
hdalbp-005	Manual	0.638	0.537	0.570	0.317	0.241	0.0235	0.1683	1.000	1.000[21]
hdalbp-006	Manual	0.867	0.480	0.786	0.420	0.000	0.0004	0.0110	1.000	1.000[22]
hdaprnu-002	Manual	0.526	0.927	0.987	0.919	0.050	0.0371	0.2877	1.000	1.000[23]
hdaprnu-004	Manual	0.985	0.440	0.695	0.309	0.000	0.0003	0.0073	1.000	1.000[24]
unibo-000	Manual	0.985	0.033	0.006	0.093	0.000	0.0001	0.0045	1.000	1.000[25]

Algorithm	Dataset	MACER*	BSCER$_q$**	BSCER$_v$** (visa)	BSCER$_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process (Bona Fides)$_v$	Failure to Process (Bona Fides)$_m$	MACER @ BSCER$_m$ =0.01	MACER @ BSCER$_m$ =0.003
idemia-004	Lincoln	0.537	-	0.002	0.005	0.000	0.0000	0.0000	0.454	0.620[1]
visteam-001	Lincoln	0.389	-	0.465	0.242	0.000	0.0000	0.0000	0.880	0.935[2]
visteam-003	Lincoln	0.704	-	0.078	0.026	0.000	0.0000	0.0000	0.880	0.935[3]
visteam-002	Lincoln	0.898	-	0.084	0.015	0.000	0.0000	0.0000	0.907	0.944[4]
visteam-004	Lincoln	0.833	-	0.044	0.020	0.000	0.0000	0.0000	0.889	0.954[5]
hhi-002	Lincoln	0.417	-	0.934	0.079	0.000	0.0000	0.0000	0.917	0.963[6]
visteam-000	Lincoln	0.648	-	0.373	0.409	0.000	0.0000	0.0000	0.944	0.972[7]
visteamicao-000	Lincoln	0.926	-	0.092	0.013	0.000	0.0000	0.0000	0.944	0.981[8]
wvusingle-002	Lincoln	0.833	-	0.138	0.028	0.000	0.0000	0.0000	0.954	0.981[9]
wvusingle-001	Lincoln	0.694	-	0.113	0.135	0.000	0.0000	0.0000	0.981	0.991[10]
neurotechnology-000	Lincoln	0.519	-	0.037	0.037	0.000	0.0023	0.0042	0.806	1.000[11]

idemia-003	Lincoln	0.917	-	0.007	0.012	0.000	0.0000	0.0000	0.926	$1.000^{(12)}$
unibo-000	Lincoln	0.694	-	0.006	0.093	0.000	0.0001	0.0045	0.935	$1.000^{(13)}$
hhi-001	Lincoln	0.417	-	0.934	0.086	0.000	0.0003	0.0074	0.972	$1.000^{(14)}$
hdabsif-004	Lincoln	0.000	-	0.001	0.711	0.000	0.0003	0.0082	1.000	$1.000^{(15)}$
hdafusion-001	Lincoln	0.694	-	0.262	0.137	0.000	0.0004	0.0124	1.000	$1.000^{(16)}$
hdafvdet-001	Lincoln	0.917	-	0.039	0.057	0.000	0.0003	0.0089	1.000	$1.000^{(17)}$
hdalaplace-001	Lincoln	0.519	-	0.003	0.177	0.000	0.0003	0.0073	1.000	$1.000^{(18)}$
hdalbp-005	Lincoln	0.796	-	0.570	0.317	0.000	0.0235	0.1683	1.000	$1.000^{(19)}$
hdalbp-006	Lincoln	0.843	-	0.786	0.420	0.000	0.0004	0.0110	1.000	$1.000^{(20)}$
hdaprnu-002	Lincoln	0.056	-	0.987	0.919	0.000	0.0371	0.2877	1.000	$1.000^{(21)}$
hdaprnu-004	Lincoln	0.917	-	0.695	0.309	0.000	0.0003	0.0073	1.000	$1.000^{(22)}$
ntnussl-002	Lincoln	1.000	-	-	0.003	0.000	-	0.0027	1.000	$1.000^{(23)}$
visionbox-000	Lincoln	0.972	-	0.253	0.023	0.000	0.0000	0.0003	1.000	$1.000^{(24)}$
visionbox-001	Lincoln	0.972	-	0.289	0.013	0.000	0.0000	0.0003	1.000	$1.000^{(25)}$

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
visionbox-000	DST	0.105	0.520	0.253	0.023	0.000	0.0000	0.0003	0.152	$0.216^{(1)}$
visionbox-001	DST	0.146	0.526	0.289	0.013	0.000	0.0000	0.0003	0.152	$0.216^{(2)}$
idemia-004	DST	0.573	0.014	0.002	0.005	0.000	0.0000	0.0000	0.532	$0.626^{(3)}$
wvusingle-002	DST	0.865	0.041	0.138	0.028	0.000	0.0000	0.0000	0.918	$0.971^{(4)}$
visteam-004	DST	0.936	0.014	0.044	0.020	0.000	0.0000	0.0000	0.959	$0.971^{(5)}$
ntnussl-002	DST	0.977	0.000	-	0.003	0.023	-	0.0027	0.977	$0.977^{(6)}$
visteam-003	DST	0.889	0.049	0.078	0.026	0.000	0.0000	0.0000	0.965	$0.982^{(7)}$
visteam-002	DST	0.965	0.023	0.084	0.015	0.000	0.0000	0.0000	0.971	$0.988^{(8)}$
wvusingle-001	DST	0.778	0.228	0.113	0.135	0.000	0.0000	0.0000	0.982	$0.988^{(9)}$
idemia-003	DST	0.825	0.000	0.007	0.012	0.000	0.0000	0.0000	0.842	$1.000^{(10)}$
hhi-002	DST	0.801	0.088	0.934	0.079	0.000	0.0000	0.0000	0.971	$1.000^{(11)}$
neurotechnology-000	DST	0.959	0.012	0.037	0.037	0.000	0.0023	0.0042	0.977	$1.000^{(12)}$
visteam-000	DST	0.661	0.310	0.373	0.409	0.000	0.0000	0.0000	0.982	$1.000^{(13)}$
visteam-001	DST	0.637	0.296	0.465	0.242	0.000	0.0000	0.0000	0.982	$1.000^{(14)}$
hdabsif-004	DST	0.035	0.916	0.001	0.711	0.000	0.0003	0.0082	1.000	$1.000^{(15)}$
hdafusion-001	DST	0.947	0.045	0.262	0.137	0.000	0.0004	0.0124	1.000	$1.000^{(16)}$
hdafvdet-001	DST	0.994	0.014	0.039	0.057	0.000	0.0003	0.0089	1.000	$1.000^{(17)}$
hdalaplace-001	DST	0.982	0.031	0.003	0.177	0.000	0.0003	0.0073	1.000	$1.000^{(18)}$
hdalbp-005	DST	0.737	0.329	0.570	0.317	0.099	0.0235	0.1683	1.000	$1.000^{(19)}$
hdalbp-006	DST	0.959	0.101	0.786	0.420	0.000	0.0004	0.0110	1.000	$1.000^{(20)}$
hdaprnu-002	DST	0.053	0.733	0.987	0.919	0.398	0.0371	0.2877	1.000	$1.000^{(21)}$
hdaprnu-004	DST	0.977	0.051	0.695	0.309	0.000	0.0003	0.0073	1.000	$1.000^{(22)}$
hhi-001	DST	0.801	0.088	0.934	0.086	0.000	0.0003	0.0074	1.000	$1.000^{(23)}$
unibo-000	DST	0.988	0.010	0.006	0.093	0.000	0.0001	0.0045	1.000	$1.000^{(24)}$
visteamicao-000	DST	1.000	0.000	0.092	0.013	0.000	0.0000	0.0000	1.000	$1.000^{(25)}$

Algorithm	Dataset	MACER	$BSCER_q$	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_q$	MACER @ $BSCER_q$=0.01	MACER @ $BSCER_q$=0.003
idemia-004	Print + Scanned	0.027	0.004	0.000	0.000	0.019	$0.041^{(1)}$
visteam-003	Print + Scanned	0.682	0.023	0.001	0.000	0.788	$0.894^{(2)}$

neurotechnology-000	Print + Scanned	0.579	0.025	0.001	0.000	0.765	$0.902^{(3)}$
visteam-004	Print + Scanned	0.830	0.007	0.001	0.000	0.814	$0.905^{(4)}$
visteamicao-000	Print + Scanned	0.667	0.025	0.001	0.000	0.819	$0.911^{(5)}$
unibo-000	Print + Scanned	0.995	0.001	0.001	0.001	0.777	$0.913^{(6)}$
wvusingle-001	Print + Scanned	0.013	0.568	0.000	0.000	0.721	$0.916^{(7)}$
visteam-002	Print + Scanned	0.839	0.008	0.001	0.000	0.833	$0.916^{(8)}$
visteam-000	Print + Scanned	0.307	0.312	0.001	0.000	0.805	$0.937^{(9)}$
hhi-002	Print + Scanned	0.767	0.020	0.000	0.000	0.867	$0.938^{(10)}$
visionbox-001	Print + Scanned	0.224	0.840	0.001	0.000	0.898	$0.952^{(11)}$
visionbox-000	Print + Scanned	0.179	0.883	0.001	0.000	0.916	$0.955^{(12)}$
visteam-001	Print + Scanned	0.392	0.258	0.001	0.000	0.925	$0.962^{(13)}$
hhi-001	Print + Scanned	0.766	0.023	0.001	0.003	0.883	$0.972^{(14)}$
wvusingle-002	Print + Scanned	0.536	0.233	0.000	0.000	0.964	$0.985^{(15)}$
ntnussl-002	Print + Scanned	0.000	0.999	0.001	0.002	0.996	$0.999^{(16)}$
idemia-003	Print + Scanned	0.099	0.022	0.000	0.000	0.163	$1.000^{(17)}$
hdafvdet-001	Print + Scanned	0.996	0.006	0.001	0.003	0.992	$1.000^{(18)}$
hdalaplace-001	Print + Scanned	0.987	0.019	0.002	0.004	0.994	$1.000^{(19)}$
hdaprnu-004	Print + Scanned	0.991	0.033	0.002	0.004	0.994	$1.000^{(20)}$
hdafusion-001	Print + Scanned	0.918	0.250	0.002	0.004	0.995	$1.000^{(21)}$
hdabsif-004	Print + Scanned	0.365	0.909	0.002	0.004	0.999	$1.000^{(22)}$
hdalbp-005	Print + Scanned	0.477	0.280	0.011	0.057	1.000	$1.000^{(23)}$
hdalbp-006	Print + Scanned	0.957	0.315	0.006	0.009	1.000	$1.000^{(24)}$
hdaprnu-002	Print + Scanned	0.459	0.796	0.002	0.038	1.000	$1.000^{(25)}$

4.3 Two-image Differential Morph Detection

Two face photos are provided to the algorithm: the first being a suspected morph and the second image representing a known, non-morphed face image of one of the subjects contributing to the morph (e.g., live capture image from an eGate). In the case that the first image is a bona fide photo, then the second image will be a known non-morphed image of the same subject taken on a different day.

4.3.1 Tier 1 - Low Quality Morphs

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
idemia-003	Online tool from website	0.031	0.051	0.009	0.025	0.000	0.0003	0.0000	0.099	$0.113^{(1)}$
secunet-004	Online tool from website	0.003	0.106	0.031	0.093	0.000	0.0006	0.0003	0.044	$0.175^{(2)}$
idemia-002	Online tool from website	0.069	0.029	0.022	0.011	0.000	0.0003	0.0000	0.077	$0.277^{(3)}$
idemia-001	Online tool from website	0.193	0.022	0.009	0.006	0.000	0.0003	0.0000	0.125	$0.411^{(4)}$
visteam-004	Online tool from website	0.511	0.039	0.121	0.055	0.002	0.0009	0.0001	0.460	$0.580^{(5)}$
secunet-003	Online tool from website	0.014	0.057	0.026	0.061	0.000	0.0023	0.0006	0.101	$0.618^{(6)}$
unibo-002	Online tool from website	0.014	0.966	0.028	0.083	0.000	0.0000	0.0000	0.333	$0.781^{(7)}$
secunet-002	Online tool from website	0.030	0.039	0.034	0.039	0.000	0.0059	0.0012	0.156	$0.820^{(8)}$
visteam-002	Online tool from website	0.408	0.043	0.068	0.054	0.002	0.0009	0.0001	0.719	$0.844^{(9)}$
visteam-003	Online tool from website	0.009	0.884	0.916	0.653	0.002	0.0009	0.0001	0.749	$0.878^{(10)}$
wvudiff-001	Online tool from website	0.056	0.527	0.654	0.417	0.005	0.0000	0.0000	0.736	$0.911^{(11)}$
visteam-001	Online tool from website	0.679	0.022	0.043	0.036	0.002	0.0009	0.0001	0.856	$0.930^{(12)}$

* MACER: This is the rate that morphs that are not detected. Lower values are better.
** BSCER: This is the rate that bona fides that were mistaken for morphs. Lower values are better.
For each dataset, the entries are ordered by the metric in the last table column.
Entries with - in them mean results are missing either due to the algorithm not being able to process the entire dataset OR results are still currently being generated.

visteamicao-000	Online tool from website	0.896	0.003	0.028	0.005	0.000	0.0000	0.0000	0.862	0.932[13]
kinit-001	Online tool from website	0.712	0.026	0.008	0.060	0.003	0.0030	0.0005	0.892	0.940[14]
visionbox-000	Online tool from website	0.024	0.194	0.162	0.102	0.001	0.0008	0.0001	0.970	0.988[15]
visionbox-001	Online tool from website	0.170	0.144	0.188	0.047	0.001	0.0008	0.0002	0.967	0.991[16]
visteam-000	Online tool from website	0.716	0.142	0.290	0.459	0.002	0.0013	0.0001	0.982	0.992[17]
ntnusub-000	Online tool from website	0.997	0.003	0.017	0.002	0.000	0.0000	0.0000	0.963	0.994[18]
ntnucan-000	Online tool from website	0.998	0.003	0.004	0.001	0.000	0.0000	0.0000	0.983	0.995[19]
hdaarcface-001	Online tool from website	0.001	0.417	0.303	0.382	0.008	0.0041	0.0039	1.000	1.000[20]
hdabsif-004	Online tool from website	0.277	0.500	0.902	0.408	0.009	0.0870	0.0108	1.000	1.000[21]
hdadfr-002	Online tool from website	0.001	0.382	0.394	0.382	0.004	0.0871	0.0116	1.000	1.000[22]
hdadfr-003	Online tool from website	0.000	0.398	0.429	0.418	0.004	0.0980	0.0127	1.000	1.000[23]
hdadfr-006	Online tool from website	0.043	0.378	0.642	0.395	0.018	0.5145	0.0101	1.000	1.000[24]
hdafusion-001	Online tool from website	0.000	0.388	0.426	0.410	0.004	0.1026	0.0143	1.000	1.000[25]
hdalaplace-001	Online tool from website	0.300	0.713	0.905	0.715	0.004	0.0870	0.0108	1.000	1.000[26]
hdalbp-006	Online tool from website	0.095	0.801	0.969	0.791	0.004	0.1006	0.0142	1.000	1.000[27]
hdamag-001	Online tool from website	0.002	0.441	0.381	0.421	0.004	0.1024	0.0140	1.000	1.000[28]
hdawl-000	Online tool from website	0.097	0.898	0.994	0.864	0.614	0.9568	0.3556	1.000	1.000[29]
hdawl-002	Online tool from website	0.193	0.758	0.884	0.833	0.004	0.1141	0.0165	1.000	1.000[30]

secunet-001	Online tool from website	0.037	0.090	0.066	0.041	0.055	0.0479	0.0150	1.000	1.000[31]
unibo-001	Online tool from website	0.003	0.989	0.088	0.147	0.000	0.0000	0.0000	1.000	1.000[32]
Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
visionbox-001	Global Morph	0.006	0.087	0.188	0.047	0.000	0.0008	0.0002	0.013	0.014[1]
visionbox-000	Global Morph	0.004	0.063	0.162	0.102	0.000	0.0008	0.0001	0.017	0.021[2]
idemia-002	Global Morph	0.119	0.000	0.022	0.011	0.000	0.0003	0.0000	0.124	0.322[3]
idemia-003	Global Morph	0.004	0.000	0.009	0.025	0.000	0.0003	0.0000	0.097	0.354[4]
secunet-004	Global Morph	0.042	0.000	0.031	0.093	0.000	0.0006	0.0003	0.214	0.449[5]
visteam-004	Global Morph	0.207	0.220	0.121	0.055	0.000	0.0009	0.0001	0.340	0.543[6]
secunet-003	Global Morph	0.064	0.000	0.026	0.061	0.000	0.0023	0.0006	0.278	0.786[7]
visteam-003	Global Morph	0.003	0.819	0.916	0.653	0.000	0.0009	0.0001	0.670	0.810[8]
secunet-002	Global Morph	0.085	0.000	0.034	0.039	0.000	0.0059	0.0012	0.261	0.843[9]
idemia-001	Global Morph	0.749	0.000	0.009	0.006	0.000	0.0003	0.0000	0.682	0.879[10]
visteam-002	Global Morph	0.460	0.031	0.068	0.054	0.000	0.0009	0.0001	0.747	0.879[11]
visteamicao-000	Global Morph	0.847	0.000	0.028	0.005	0.000	0.0000	0.0000	0.767	0.897[12]
kinit-001	Global Morph	0.631	0.087	0.008	0.060	0.000	0.0030	0.0005	0.867	0.939[13]
visteam-001	Global Morph	0.638	0.008	0.043	0.036	0.000	0.0009	0.0001	0.858	0.947[14]
wvudiff-001	Global Morph	0.089	0.087	0.654	0.417	0.000	0.0000	0.0000	0.899	0.971[15]
visteam-000	Global Morph	0.395	0.488	0.290	0.459	0.000	0.0013	0.0001	0.934	0.981[16]
ntnusub-000	Global Morph	1.000	0.000	0.017	0.002	0.000	0.0000	0.0000	0.990	1.000[17]
hdaarcface-001	Global Morph	0.026	0.031	0.303	0.382	0.010	0.0041	0.0039	1.000	1.000[18]
hdabsif-004	Global Morph	0.558	0.165	0.902	0.408	0.000	0.0870	0.0108	1.000	1.000[19]
hdadfr-002	Global Morph	0.025	0.024	0.394	0.382	0.000	0.0871	0.0116	1.000	1.000[20]
hdadfr-003	Global Morph	0.027	0.016	0.429	0.418	0.000	0.0980	0.0127	1.000	1.000[21]
hdadfr-006	Global Morph	0.199	0.039	0.642	0.395	0.000	0.5145	0.0101	1.000	1.000[22]

hdafusion-001	Global Morph	0.012	0.047	0.426	0.410	0.000	0.1026	0.0143	1.000	1.000[23]
hdalaplace-001	Global Morph	0.242	0.606	0.905	0.715	0.000	0.0870	0.0108	1.000	1.000[24]
hdalbp-006	Global Morph	0.149	0.504	0.969	0.791	0.000	0.1006	0.0142	1.000	1.000[25]
hdamag-001	Global Morph	0.019	0.016	0.381	0.421	0.000	0.1024	0.0140	1.000	1.000[26]
hdawl-000	Global Morph	0.288	0.614	0.994	0.864	0.134	0.9568	0.3556	1.000	1.000[27]
hdawl-002	Global Morph	0.207	0.496	0.884	0.833	0.000	0.1141	0.0165	1.000	1.000[28]
ntnucan-000	Global Morph	1.000	0.000	0.004	0.001	0.000	0.0000	0.0000	1.000	1.000[29]
secunet-001	Global Morph	0.103	0.000	0.066	0.041	0.000	0.0479	0.0150	1.000	1.000[30]
unibo-001	Global Morph	0.957	0.047	0.088	0.147	0.000	0.0000	0.0000	1.000	1.000[31]
unibo-002	Global Morph	0.974	0.031	0.028	0.083	0.000	0.0000	0.0000	1.000	1.000[32]

4.3.2 Tier 2 - Automated Morphs

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
visionbox-001	Local Morph	0.003	0.087	0.188	0.047	0.000	0.0008	0.0002	0.006	0.008[1]
visionbox-000	Local Morph	0.001	0.063	0.162	0.102	0.000	0.0008	0.0001	0.006	0.009[2]
idemia-003	Local Morph	0.009	0.000	0.009	0.025	0.000	0.0003	0.0000	0.074	0.264[3]
idemia-002	Local Morph	0.072	0.000	0.022	0.011	0.000	0.0003	0.0000	0.080	0.280[4]
secunet-004	Local Morph	0.022	0.000	0.031	0.093	0.000	0.0006	0.0003	0.158	0.389[5]
visteam-004	Local Morph	0.165	0.220	0.121	0.055	0.000	0.0009	0.0001	0.290	0.482[6]
visteam-003	Local Morph	0.001	0.819	0.916	0.653	0.000	0.0009	0.0001	0.601	0.756[7]
secunet-003	Local Morph	0.043	0.000	0.026	0.061	0.000	0.0023	0.0006	0.241	0.779[8]
visteamicao-000	Local Morph	0.727	0.000	0.028	0.005	0.000	0.0000	0.0000	0.638	0.809[9]
visteam-002	Local Morph	0.336	0.031	0.068	0.054	0.000	0.0009	0.0001	0.662	0.814[10]
idemia-001	Local Morph	0.669	0.000	0.009	0.006	0.000	0.0003	0.0000	0.581	0.828[11]
secunet-002	Local Morph	0.065	0.000	0.034	0.039	0.000	0.0059	0.0012	0.235	0.843[12]
visteam-001	Local Morph	0.553	0.008	0.043	0.036	0.000	0.0009	0.0001	0.789	0.912[13]
visteam-000	Local Morph	0.358	0.488	0.290	0.459	0.000	0.0013	0.0001	0.918	0.971[14]
wvudiff-001	Local Morph	0.083	0.087	0.654	0.417	0.000	0.0000	0.0000	0.913	0.974[15]
kinit-001	Local Morph	0.823	0.087	0.008	0.060	0.000	0.0030	0.0005	0.959	0.985[16]
ntnucan-000	Local Morph	1.000	0.000	0.004	0.001	0.000	0.0000	0.0000	0.997	0.999[17]
ntnusub-000	Local Morph	1.000	0.000	0.017	0.002	0.000	0.0000	0.0000	0.990	1.000[18]
hdaarcface-001	Local Morph	0.016	0.031	0.303	0.382	0.010	0.0041	0.0039	1.000	1.000[19]
hdabsif-004	Local Morph	0.596	0.165	0.902	0.408	0.000	0.0870	0.0108	1.000	1.000[20]
hdadfr-002	Local Morph	0.017	0.024	0.394	0.382	0.000	0.0871	0.0116	1.000	1.000[21]
hdadfr-003	Local Morph	0.014	0.016	0.429	0.418	0.001	0.0980	0.0127	1.000	1.000[22]
hdadfr-006	Local Morph	0.159	0.039	0.642	0.395	0.000	0.5145	0.0101	1.000	1.000[23]

hdafusion-001	Local Morph	0.007	0.047	0.426	0.410	0.001	0.1026	0.0143	1.000	1.000[24]
hdalaplace-001	Local Morph	0.230	0.606	0.905	0.715	0.000	0.0870	0.0108	1.000	1.000[25]
hdalbp-006	Local Morph	0.134	0.504	0.969	0.791	0.001	0.1006	0.0142	1.000	1.000[26]
hdamag-001	Local Morph	0.017	0.016	0.381	0.421	0.001	0.1024	0.0140	1.000	1.000[27]
hdawl-000	Local Morph	0.230	0.614	0.994	0.864	0.165	0.9568	0.3556	1.000	1.000[28]
hdawl-002	Local Morph	0.175	0.496	0.884	0.833	0.001	0.1141	0.0165	1.000	1.000[29]
secunet-001	Local Morph	0.084	0.000	0.066	0.041	0.000	0.0479	0.0150	1.000	1.000[30]
unibo-001	Local Morph	0.959	0.047	0.088	0.147	0.000	0.0000	0.0000	1.000	1.000[31]
unibo-002	Local Morph	0.970	0.031	0.028	0.083	0.000	0.0000	0.0000	1.000	1.000[32]

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
visionbox-001	Local Morph Colorized Average	0.003	0.087	0.188	0.047	0.000	0.0008	0.0002	0.009	0.011[1]
visionbox-000	Local Morph Colorized Average	0.001	0.063	0.162	0.102	0.000	0.0008	0.0001	0.013	0.020[2]
idemia-003	Local Morph Colorized Average	0.027	0.000	0.009	0.025	0.000	0.0003	0.0000	0.106	0.288[3]
idemia-002	Local Morph Colorized Average	0.163	0.000	0.022	0.011	0.000	0.0003	0.0000	0.173	0.390[4]
secunet-004	Local Morph Colorized Average	0.031	0.000	0.031	0.093	0.000	0.0006	0.0003	0.179	0.416[5]
visteam-004	Local Morph Colorized Average	0.267	0.220	0.121	0.055	0.000	0.0009	0.0001	0.374	0.549[6]
secunet-003	Local Morph Colorized Average	0.057	0.000	0.026	0.061	0.000	0.0023	0.0006	0.263	0.799[7]
visteam-003	Local Morph Colorized Average	0.003	0.819	0.916	0.653	0.000	0.0009	0.0001	0.678	0.809[8]
idemia-001	Local Morph Colorized Average	0.671	0.000	0.009	0.006	0.000	0.0003	0.0000	0.582	0.830[9]
secunet-002	Local Morph Colorized Average	0.076	0.000	0.034	0.039	0.000	0.0059	0.0012	0.251	0.854[10]
visteamicao-000	Local Morph Colorized Average	0.793	0.000	0.028	0.005	0.000	0.0000	0.0000	0.708	0.858[11]
visteam-002	Local Morph Colorized Average	0.391	0.031	0.068	0.054	0.000	0.0009	0.0001	0.724	0.868[12]
visteam-001	Local Morph Colorized Average	0.593	0.008	0.043	0.036	0.000	0.0009	0.0001	0.839	0.938[13]

wvudiff-001	Local Morph Colorized Average	0.080	0.087	0.654	0.417	0.000	0.0000	0.0000	0.911	0.976[14]
visteam-000	Local Morph Colorized Average	0.386	0.488	0.290	0.459	0.000	0.0013	0.0001	0.933	0.978[15]
kinit-001	Local Morph Colorized Average	0.850	0.087	0.008	0.060	0.000	0.0030	0.0005	0.968	0.992[16]
ntnusub-000	Local Morph Colorized Average	1.000	0.000	0.017	0.002	0.000	0.0000	0.0000	0.991	1.000[17]
hdaarcface-001	Local Morph Colorized Average	0.016	0.031	0.303	0.382	0.011	0.0041	0.0039	1.000	1.000[18]
hdabsif-004	Local Morph Colorized Average	0.602	0.165	0.902	0.408	0.000	0.0870	0.0108	1.000	1.000[19]
hdadfr-002	Local Morph Colorized Average	0.018	0.024	0.394	0.382	0.000	0.0871	0.0116	1.000	1.000[20]
hdadfr-003	Local Morph Colorized Average	0.015	0.016	0.429	0.418	0.001	0.0980	0.0127	1.000	1.000[21]
hdadfr-006	Local Morph Colorized Average	0.161	0.039	0.642	0.395	0.000	0.5145	0.0101	1.000	1.000[22]
hdafusion-001	Local Morph Colorized Average	0.007	0.047	0.426	0.410	0.001	0.1026	0.0143	1.000	1.000[23]
hdalaplace-001	Local Morph Colorized Average	0.232	0.606	0.905	0.715	0.000	0.0870	0.0108	1.000	1.000[24]
hdalbp-006	Local Morph Colorized Average	0.134	0.504	0.969	0.791	0.000	0.1006	0.0142	1.000	1.000[25]
hdamag-001	Local Morph Colorized Average	0.016	0.016	0.381	0.421	0.001	0.1024	0.0140	1.000	1.000[26]
hdawl-000	Local Morph Colorized Average	0.257	0.614	0.994	0.864	0.147	0.9568	0.3556	1.000	1.000[27]
hdawl-002	Local Morph Colorized Average	0.182	0.496	0.884	0.833	0.001	0.1141	0.0165	1.000	1.000[28]
ntnucan-000	Local Morph Colorized Average	1.000	0.000	0.004	0.001	0.000	0.0000	0.0000	1.000	1.000[29]
secunet-001	Local Morph Colorized Average	0.100	0.000	0.066	0.041	0.000	0.0479	0.0150	1.000	1.000[30]
unibo-001	Local Morph Colorized Average	0.953	0.047	0.088	0.147	0.000	0.0000	0.0000	1.000	1.000[31]

unibo-002	Local Morph Colorized Average	0.969	0.031	0.028	0.083	0.000	0.0000	0.0000	1.000	1.000[32]

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
visionbox-001	Local Morph Colorized Match	0.006	0.087	0.188	0.047	0.000	0.0008	0.0002	0.014	0.022[1]
visionbox-000	Local Morph Colorized Match	0.008	0.063	0.162	0.102	0.000	0.0008	0.0001	0.018	0.028[2]
idemia-003	Local Morph Colorized Match	0.049	0.000	0.009	0.025	0.000	0.0003	0.0000	0.135	0.314[3]
secunet-004	Local Morph Colorized Match	0.031	0.000	0.031	0.093	0.000	0.0006	0.0003	0.179	0.427[4]
idemia-002	Local Morph Colorized Match	0.206	0.000	0.022	0.011	0.000	0.0003	0.0000	0.220	0.454[5]
visteam-004	Local Morph Colorized Match	0.328	0.220	0.121	0.055	0.000	0.0009	0.0001	0.437	0.598[6]
secunet-003	Local Morph Colorized Match	0.062	0.000	0.026	0.061	0.000	0.0023	0.0006	0.268	0.805[7]
idemia-001	Local Morph Colorized Match	0.649	0.000	0.009	0.006	0.000	0.0003	0.0000	0.565	0.815[8]
visteam-003	Local Morph Colorized Match	0.003	0.819	0.916	0.653	0.000	0.0009	0.0001	0.735	0.862[9]
secunet-002	Local Morph Colorized Match	0.087	0.000	0.034	0.039	0.000	0.0059	0.0012	0.266	0.865[10]
visteamicao-000	Local Morph Colorized Match	0.868	0.000	0.028	0.005	0.000	0.0000	0.0000	0.811	0.916[11]
visteam-002	Local Morph Colorized Match	0.545	0.031	0.068	0.054	0.000	0.0009	0.0001	0.816	0.918[12]
visteam-001	Local Morph Colorized Match	0.694	0.008	0.043	0.036	0.000	0.0009	0.0001	0.878	0.957[13]
visteam-000	Local Morph Colorized Match	0.427	0.488	0.290	0.459	0.000	0.0013	0.0001	0.949	0.983[14]
kinit-001	Local Morph Colorized Match	0.863	0.087	0.008	0.060	0.000	0.0030	0.0005	0.977	0.990[15]
wvudiff-001	Local Morph Colorized Match	0.306	0.087	0.654	0.417	0.000	0.0000	0.0000	0.971	0.991[16]

ntnucan-000	Local Morph Colorized Match	1.000	0.000	0.004	0.001	0.000	0.0000	0.0000	0.997	0.998[17]
ntnusub-000	Local Morph Colorized Match	1.000	0.000	0.017	0.002	0.000	0.0000	0.0000	0.990	1.000[18]
hdaarcface-001	Local Morph Colorized Match	0.016	0.031	0.303	0.382	0.010	0.0041	0.0039	1.000	1.000[19]
hdabsif-004	Local Morph Colorized Match	0.616	0.165	0.902	0.408	0.000	0.0870	0.0108	1.000	1.000[20]
hdadfr-002	Local Morph Colorized Match	0.021	0.024	0.394	0.382	0.000	0.0871	0.0116	1.000	1.000[21]
hdadfr-003	Local Morph Colorized Match	0.019	0.016	0.429	0.418	0.001	0.0980	0.0127	1.000	1.000[22]
hdadfr-006	Local Morph Colorized Match	0.183	0.039	0.642	0.395	0.000	0.5145	0.0101	1.000	1.000[23]
hdafusion-001	Local Morph Colorized Match	0.011	0.047	0.426	0.410	0.001	0.1026	0.0143	1.000	1.000[24]
hdalaplace-001	Local Morph Colorized Match	0.286	0.606	0.905	0.715	0.000	0.0870	0.0108	1.000	1.000[25]
hdalbp-006	Local Morph Colorized Match	0.155	0.504	0.969	0.791	0.000	0.1006	0.0142	1.000	1.000[26]
hdamag-001	Local Morph Colorized Match	0.017	0.016	0.381	0.421	0.001	0.1024	0.0140	1.000	1.000[27]
hdawl-000	Local Morph Colorized Match	0.232	0.614	0.994	0.864	0.155	0.9568	0.3556	1.000	1.000[28]
hdawl-002	Local Morph Colorized Match	0.191	0.496	0.884	0.833	0.001	0.1141	0.0165	1.000	1.000[29]
secunet-001	Local Morph Colorized Match	0.106	0.000	0.066	0.041	0.000	0.0479	0.0150	1.000	1.000[30]
unibo-001	Local Morph Colorized Match	0.956	0.047	0.088	0.147	0.000	0.0000	0.0000	1.000	1.000[31]
unibo-002	Local Morph Colorized Match	0.967	0.031	0.028	0.083	0.000	0.0000	0.0000	1.000	1.000[32]

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003

visionbox-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.001	0.141	0.188	0.047	0.000	0.0008	0.0002	0.002	$0.003^{(1)}$
visionbox-000	UNIBO Automatic Morphed Face Generation Tool v1.0	0.001	0.141	0.162	0.102	0.000	0.0008	0.0001	0.004	$0.006^{(2)}$
idemia-003	UNIBO Automatic Morphed Face Generation Tool v1.0	0.001	0.031	0.009	0.025	0.000	0.0003	0.0000	0.020	$0.116^{(3)}$
visteamicao-000	UNIBO Automatic Morphed Face Generation Tool v1.0	0.111	0.000	0.028	0.005	0.000	0.0000	0.0000	0.079	$0.149^{(4)}$
idemia-002	UNIBO Automatic Morphed Face Generation Tool v1.0	0.054	0.031	0.022	0.011	0.000	0.0003	0.0000	0.059	$0.168^{(5)}$
secunet-004	UNIBO Automatic Morphed Face Generation Tool v1.0	0.006	0.031	0.031	0.093	0.000	0.0006	0.0003	0.081	$0.241^{(6)}$
idemia-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.371	0.016	0.009	0.006	0.000	0.0003	0.0000	0.320	$0.505^{(7)}$
visteam-004	UNIBO Automatic Morphed Face Generation Tool v1.0	0.214	0.016	0.121	0.055	0.000	0.0009	0.0001	0.312	$0.525^{(8)}$
secunet-003	UNIBO Automatic Morphed Face Generation Tool v1.0	0.015	0.016	0.026	0.061	0.000	0.0023	0.0006	0.136	$0.609^{(9)}$

MACER(T) Morph Miss Rate
BSCER(T) False Detection Rate

secunet-002	UNIBO Automatic Morphed Face Generation Tool v1.0	0.012	0.031	0.034	0.039	0.028	0.0059	0.0012	0.087	0.672[10]
visteam-002	UNIBO Automatic Morphed Face Generation Tool v1.0	0.190	0.047	0.068	0.054	0.000	0.0009	0.0001	0.512	0.726[11]
visteam-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.278	0.047	0.043	0.036	0.000	0.0009	0.0001	0.551	0.768[12]
visteam-003	UNIBO Automatic Morphed Face Generation Tool v1.0	0.004	0.844	0.916	0.653	0.000	0.0009	0.0001	0.668	0.844[13]
wvudiff-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.063	0.438	0.654	0.417	0.000	0.0000	0.0000	0.733	0.933[14]
visteam-000	UNIBO Automatic Morphed Face Generation Tool v1.0	0.392	0.438	0.290	0.459	0.000	0.0013	0.0001	0.887	0.956[15]
ntnusub-000	UNIBO Automatic Morphed Face Generation Tool v1.0	0.972	0.031	0.017	0.002	0.000	0.0000	0.0000	0.955	0.966[16]
ntnucan-000	UNIBO Automatic Morphed Face Generation Tool v1.0	0.996	0.016	0.004	0.001	0.000	0.0000	0.0000	0.959	0.967[17]
kinit-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.966	0.016	0.008	0.060	0.000	0.0030	0.0005	0.999	1.000[18]

hdaarcface-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.008	0.188	0.303	0.382	0.006	0.0041	0.0039	1.000	1.000[19]
hdabsif-004	UNIBO Automatic Morphed Face Generation Tool v1.0	0.160	0.750	0.902	0.408	0.034	0.0870	0.0108	1.000	1.000[20]
hdadfr-002	UNIBO Automatic Morphed Face Generation Tool v1.0	0.008	0.219	0.394	0.382	0.034	0.0871	0.0116	1.000	1.000[21]
hdadfr-003	UNIBO Automatic Morphed Face Generation Tool v1.0	0.008	0.234	0.429	0.418	0.034	0.0980	0.0127	1.000	1.000[22]
hdadfr-006	UNIBO Automatic Morphed Face Generation Tool v1.0	0.031	0.406	0.642	0.395	0.037	0.5145	0.0101	1.000	1.000[23]
hdafusion-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.002	0.312	0.426	0.410	0.034	0.1026	0.0143	1.000	1.000[24]
hdalaplace-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.294	0.734	0.905	0.715	0.034	0.0870	0.0108	1.000	1.000[25]
hdalbp-006	UNIBO Automatic Morphed Face Generation Tool v1.0	0.001	0.969	0.969	0.791	0.034	0.1006	0.0142	1.000	1.000[26]
hdamag-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.006	0.266	0.381	0.421	0.034	0.1024	0.0140	1.000	1.000[27]

hdawl-000	UNIBO Automatic Morphed Face Generation Tool v1.0	0.029	0.984	0.994	0.864	0.758	0.9568	0.3556	1.000	1.000[28]
hdawl-002	UNIBO Automatic Morphed Face Generation Tool v1.0	0.063	0.938	0.884	0.833	0.034	0.1141	0.0165	1.000	1.000[29]
secunet-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.018	0.125	0.066	0.041	0.140	0.0479	0.0150	1.000	1.000[30]
unibo-001	UNIBO Automatic Morphed Face Generation Tool v1.0	0.882	0.047	0.088	0.147	0.000	0.0000	0.0000	1.000	1.000[31]
unibo-002	UNIBO Automatic Morphed Face Generation Tool v1.0	0.943	0.016	0.028	0.083	0.000	0.0000	0.0000	1.000	1.000[32]

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
visionbox-001	Visa-Border	0.002	0.138	0.188	0.047	0.0002	0.0008	0.0002	0.003	0.004[1]
visionbox-000	Visa-Border	0.001	0.115	0.162	0.102	0.0002	0.0008	0.0001	0.004	0.005[2]
idemia-003	Visa-Border	0.016	0.005	0.009	0.025	0.0002	0.0003	0.0000	0.040	0.111[3]
idemia-002	Visa-Border	0.031	0.011	0.022	0.011	0.0002	0.0003	0.0000	0.035	0.148[4]
secunet-004	Visa-Border	0.020	0.017	0.031	0.093	0.0002	0.0006	0.0003	0.179	0.444[5]
visteamicao-000	Visa-Border	0.460	0.023	0.028	0.005	0.0000	0.0000	0.0000	0.388	0.535[6]
idemia-001	Visa-Border	0.401	0.003	0.009	0.006	0.0002	0.0003	0.0000	0.337	0.585[7]
visteam-004	Visa-Border	0.556	0.116	0.121	0.055	0.0002	0.0009	0.0001	0.624	0.775[8]
visteam-003	Visa-Border	0.011	0.921	0.916	0.653	0.0002	0.0009	0.0001	0.682	0.822[9]
secunet-003	Visa-Border	0.041	0.014	0.026	0.061	0.0007	0.0023	0.0006	0.256	0.847[10]
secunet-002	Visa-Border	0.048	0.019	0.034	0.039	0.0027	0.0059	0.0012	0.212	0.888[11]
visteam-002	Visa-Border	0.513	0.053	0.068	0.054	0.0002	0.0009	0.0001	0.815	0.928[12]
visteam-001	Visa-Border	0.659	0.032	0.043	0.036	0.0002	0.0009	0.0001	0.844	0.936[13]
wvudiff-001	Visa-Border	0.123	0.609	0.654	0.417	0.0000	0.0000	0.0000	0.901	0.971[14]
ntnusub-000	Visa-Border	0.987	0.013	0.017	0.002	0.0000	0.0000	0.0000	0.939	0.983[15]
visteam-000	Visa-Border	0.623	0.277	0.290	0.459	0.0003	0.0013	0.0001	0.967	0.984[16]
ntnucan-000	Visa-Border	0.998	0.002	0.004	0.001	0.0000	0.0000	0.0000	0.967	0.986[17]
kinit-001	Visa-Border	0.999	0.004	0.008	0.060	0.0011	0.0030	0.0005	0.999	0.999[18]
unibo-002	Visa-Border	0.977	0.038	0.028	0.083	0.0000	0.0000	0.0000	0.999	0.999[19]

hdaarcface-001	Visa-Border	0.008	0.261	0.303	0.382	0.0017	0.0041	0.0039	1.000	1.000[20]
hdabsif-004	Visa-Border	0.218	0.859	0.902	0.408	0.0679	0.0870	0.0108	1.000	1.000[21]
hdadfr-002	Visa-Border	0.006	0.349	0.394	0.382	0.0680	0.0871	0.0116	1.000	1.000[22]
hdadfr-003	Visa-Border	0.005	0.383	0.429	0.418	0.0764	0.0980	0.0127	1.000	1.000[23]
hdadfr-006	Visa-Border	0.044	0.604	0.642	0.395	0.4911	0.5145	0.0101	1.000	1.000[24]
hdafusion-001	Visa-Border	0.005	0.367	0.426	0.410	0.0812	0.1026	0.0143	1.000	1.000[25]
hdalaplace-001	Visa-Border	0.038	0.913	0.905	0.715	0.0679	0.0870	0.0108	1.000	1.000[26]
hdalbp-006	Visa-Border	0.001	0.965	0.969	0.791	0.0791	0.1006	0.0142	1.000	1.000[27]
hdamag-001	Visa-Border	0.005	0.333	0.381	0.421	0.0809	0.1024	0.0140	1.000	1.000[28]
hdawl-000	Visa-Border	0.005	0.993	0.994	0.864	0.9514	0.9568	0.3556	1.000	1.000[29]
hdawl-002	Visa-Border	0.089	0.864	0.884	0.833	0.0894	0.1141	0.0165	1.000	1.000[30]
secunet-001	Visa-Border	0.061	0.045	0.066	0.041	0.0343	0.0479	0.0150	1.000	1.000[31]
unibo-001	Visa-Border	0.924	0.106	0.088	0.147	0.0000	0.0000	0.0000	1.000	1.000[32]

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
visionbox-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.002	0.141	0.188	0.047	0.000	0.0008	0.0002	0.006	0.010[1]
visionbox-000	UNIBO Automatic Morphed Face Generation Tool v2.0	0.002	0.141	0.162	0.102	0.000	0.0008	0.0001	0.008	0.013[2]
idemia-003	UNIBO Automatic Morphed Face Generation Tool v2.0	0.004	0.031	0.009	0.025	0.000	0.0003	0.0000	0.038	0.143[3]
idemia-002	UNIBO Automatic Morphed Face Generation Tool v2.0	0.077	0.031	0.022	0.011	0.000	0.0003	0.0000	0.083	0.216[4]
secunet-004	UNIBO Automatic Morphed Face Generation Tool v2.0	0.012	0.031	0.031	0.093	0.000	0.0006	0.0003	0.103	0.271[5]
visteamicao-000	UNIBO Automatic Morphed Face Generation Tool v2.0	0.227	0.000	0.028	0.005	0.000	0.0000	0.0000	0.163	0.305[6]

idemia-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.379	0.016	0.009	0.006	0.000	0.0003	0.0000	0.334	0.487$^{(7)}$
visteam-004	UNIBO Automatic Morphed Face Generation Tool v2.0	0.320	0.016	0.121	0.055	0.000	0.0009	0.0001	0.392	0.591$^{(8)}$
secunet-003	UNIBO Automatic Morphed Face Generation Tool v2.0	0.023	0.016	0.026	0.061	0.000	0.0023	0.0006	0.160	0.630$^{(9)}$
secunet-002	UNIBO Automatic Morphed Face Generation Tool v2.0	0.021	0.031	0.034	0.039	0.028	0.0059	0.0012	0.112	0.701$^{(10)}$
visteam-002	UNIBO Automatic Morphed Face Generation Tool v2.0	0.275	0.047	0.068	0.054	0.000	0.0009	0.0001	0.630	0.809$^{(11)}$
visteam-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.379	0.047	0.043	0.036	0.000	0.0009	0.0001	0.648	0.837$^{(12)}$
visteam-003	UNIBO Automatic Morphed Face Generation Tool v2.0	0.004	0.844	0.916	0.653	0.000	0.0009	0.0001	0.747	0.894$^{(13)}$
visteam-000	UNIBO Automatic Morphed Face Generation Tool v2.0	0.415	0.438	0.290	0.459	0.000	0.0013	0.0001	0.890	0.961$^{(14)}$
wvudiff-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.075	0.438	0.654	0.417	0.000	0.0000	0.0000	0.831	0.965$^{(15)}$

ntnusub-000	UNIBO Automatic Morphed Face Generation Tool v2.0	0.972	0.031	0.017	0.002	0.000	0.0000	0.0000	0.954	0.966[16]
ntnucan-000	UNIBO Automatic Morphed Face Generation Tool v2.0	0.996	0.016	0.004	0.001	0.000	0.0000	0.0000	0.958	0.967[17]
kinit-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.970	0.016	0.008	0.060	0.000	0.0030	0.0005	0.998	1.000[18]
hdaarcface-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.009	0.188	0.303	0.382	0.006	0.0041	0.0039	1.000	1.000[19]
hdabsif-004	UNIBO Automatic Morphed Face Generation Tool v2.0	0.146	0.750	0.902	0.408	0.034	0.0870	0.0108	1.000	1.000[20]
hdadfr-002	UNIBO Automatic Morphed Face Generation Tool v2.0	0.009	0.219	0.394	0.382	0.034	0.0871	0.0116	1.000	1.000[21]
hdadfr-003	UNIBO Automatic Morphed Face Generation Tool v2.0	0.010	0.234	0.429	0.418	0.034	0.0980	0.0127	1.000	1.000[22]
hdadfr-006	UNIBO Automatic Morphed Face Generation Tool v2.0	0.040	0.406	0.642	0.395	0.037	0.5145	0.0101	1.000	1.000[23]
hdafusion-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.003	0.312	0.426	0.410	0.034	0.1026	0.0143	1.000	1.000[24]

hdalaplace-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.281	0.734	0.905	0.715	0.034	0.0870	0.0108	1.000	1.000[25]
hdalbp-006	UNIBO Automatic Morphed Face Generation Tool v2.0	0.001	0.969	0.969	0.791	0.034	0.1006	0.0142	1.000	1.000[26]
hdamag-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.006	0.266	0.381	0.421	0.034	0.1024	0.0140	1.000	1.000[27]
hdawl-000	UNIBO Automatic Morphed Face Generation Tool v2.0	0.030	0.984	0.994	0.864	0.758	0.9568	0.3556	1.000	1.000[28]
hdawl-002	UNIBO Automatic Morphed Face Generation Tool v2.0	0.062	0.938	0.884	0.833	0.034	0.1141	0.0165	1.000	1.000[29]
secunet-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.028	0.125	0.066	0.041	0.140	0.0479	0.0150	1.000	1.000[30]
unibo-001	UNIBO Automatic Morphed Face Generation Tool v2.0	0.877	0.047	0.088	0.147	0.000	0.0000	0.0000	1.000	1.000[31]
unibo-002	UNIBO Automatic Morphed Face Generation Tool v2.0	0.942	0.016	0.028	0.083	0.000	0.0000	0.0000	1.000	1.000[32]

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
visionbox-001	Twente	0.000	0.141	0.188	0.047	0.000	0.0008	0.0002	0.002	0.003[1]
visionbox-000	Twente	0.001	0.141	0.162	0.102	0.000	0.0008	0.0001	0.004	0.006[2]
idemia-003	Twente	0.000	0.031	0.009	0.025	0.000	0.0003	0.0000	0.020	0.102[3]
idemia-002	Twente	0.062	0.031	0.022	0.011	0.000	0.0003	0.0000	0.065	0.190[4]

MACER(T) Morph Miss Rate
BSCER(T) False Detection Rate

visteamicao-000	Twente	0.136	0.000	0.028	0.005	0.000	0.0000	0.0000	0.094	0.194$^{(5)}$
secunet-004	Twente	0.010	0.031	0.031	0.093	0.000	0.0006	0.0003	0.084	0.245$^{(6)}$
idemia-001	Twente	0.361	0.016	0.009	0.006	0.000	0.0003	0.0000	0.313	0.485$^{(7)}$
visteam-004	Twente	0.268	0.016	0.121	0.055	0.000	0.0009	0.0001	0.335	0.537$^{(8)}$
secunet-003	Twente	0.017	0.016	0.026	0.061	0.000	0.0023	0.0006	0.134	0.603$^{(9)}$
secunet-002	Twente	0.016	0.031	0.034	0.039	0.028	0.0059	0.0012	0.102	0.697$^{(10)}$
visteam-002	Twente	0.286	0.047	0.068	0.054	0.000	0.0009	0.0001	0.615	0.803$^{(11)}$
visteam-001	Twente	0.372	0.047	0.043	0.036	0.000	0.0009	0.0001	0.644	0.840$^{(12)}$
visteam-003	Twente	0.005	0.844	0.916	0.653	0.000	0.0009	0.0001	0.771	0.898$^{(13)}$
wvudiff-001	Twente	0.062	0.438	0.654	0.417	0.000	0.0000	0.0000	0.758	0.943$^{(14)}$
visteam-000	Twente	0.425	0.438	0.290	0.459	0.000	0.0013	0.0001	0.887	0.958$^{(15)}$
ntnusub-000	Twente	0.972	0.031	0.017	0.002	0.000	0.0000	0.0000	0.955	0.966$^{(16)}$
ntnucan-000	Twente	0.995	0.016	0.004	0.001	0.000	0.0000	0.0000	0.960	0.969$^{(17)}$
kinit-001	Twente	0.911	0.016	0.008	0.060	0.000	0.0030	0.0005	0.990	0.999$^{(18)}$
hdaarcface-001	Twente	0.008	0.188	0.303	0.382	0.006	0.0041	0.0039	1.000	1.000$^{(19)}$
hdabsif-004	Twente	0.255	0.750	0.902	0.408	0.034	0.0870	0.0108	1.000	1.000$^{(20)}$
hdadfr-002	Twente	0.007	0.219	0.394	0.382	0.034	0.0871	0.0116	1.000	1.000$^{(21)}$
hdadfr-003	Twente	0.008	0.234	0.429	0.418	0.034	0.0980	0.0127	1.000	1.000$^{(22)}$
hdadfr-006	Twente	0.040	0.406	0.642	0.395	0.037	0.5145	0.0101	1.000	1.000$^{(23)}$
hdafusion-001	Twente	0.003	0.312	0.426	0.410	0.034	0.1026	0.0143	1.000	1.000$^{(24)}$
hdalaplace-001	Twente	0.242	0.734	0.905	0.715	0.034	0.0870	0.0108	1.000	1.000$^{(25)}$
hdalbp-006	Twente	0.002	0.969	0.969	0.791	0.034	0.1006	0.0142	1.000	1.000$^{(26)}$
hdamag-001	Twente	0.006	0.266	0.381	0.421	0.034	0.1024	0.0140	1.000	1.000$^{(27)}$
hdawl-000	Twente	0.028	0.984	0.994	0.864	0.758	0.9568	0.3556	1.000	1.000$^{(28)}$
hdawl-002	Twente	0.067	0.938	0.884	0.833	0.034	0.1141	0.0165	1.000	1.000$^{(29)}$
secunet-001	Twente	0.022	0.125	0.066	0.041	0.140	0.0479	0.0150	1.000	1.000$^{(30)}$
unibo-001	Twente	0.948	0.047	0.088	0.147	0.000	0.0000	0.0000	1.000	1.000$^{(31)}$
unibo-002	Twente	0.985	0.016	0.028	0.083	0.000	0.0000	0.0000	1.000	1.000$^{(32)}$

Algorithm	Dataset	MACER*	BSCER$_q$**	BSCER$_v$** (visa)	BSCER$_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process (Bona Fides)$_v$	Failure to Process (Bona Fides)$_m$	MACER @ BSCER$_m$ =0.01	MACER @ BSCER$_m$ =0.003
idemia-003	MIPGAN-II	0.000	0.031	0.009	0.025	0.000	0.0003	0.0000	0.000	0.000$^{(1)}$
visionbox-001	MIPGAN-II	0.002	0.141	0.188	0.047	0.000	0.0008	0.0002	0.003	0.006$^{(2)}$
visionbox-000	MIPGAN-II	0.001	0.141	0.162	0.102	0.000	0.0008	0.0001	0.006	0.011$^{(3)}$
idemia-002	MIPGAN-II	0.002	0.031	0.022	0.011	0.000	0.0003	0.0000	0.003	0.037$^{(4)}$
idemia-001	MIPGAN-II	0.017	0.016	0.009	0.006	0.000	0.0003	0.0000	0.006	0.108$^{(5)}$
secunet-004	MIPGAN-II	0.004	0.031	0.031	0.093	0.000	0.0006	0.0003	0.062	0.285$^{(6)}$
wvudiff-001	MIPGAN-II	0.019	0.438	0.654	0.417	0.000	0.0000	0.0000	0.481	0.714$^{(7)}$
secunet-003	MIPGAN-II	0.012	0.016	0.026	0.061	0.000	0.0023	0.0006	0.147	0.744$^{(8)}$
visteam-004	MIPGAN-II	0.669	0.016	0.121	0.055	0.000	0.0009	0.0001	0.631	0.791$^{(9)}$
secunet-002	MIPGAN-II	0.024	0.031	0.034	0.039	0.028	0.0059	0.0012	0.134	0.860$^{(10)}$
visteamicao-000	MIPGAN-II	0.880	0.000	0.028	0.005	0.000	0.0000	0.0000	0.818	0.922$^{(11)}$
visteam-002	MIPGAN-II	0.530	0.047	0.068	0.054	0.000	0.0009	0.0001	0.859	0.959$^{(12)}$
visteam-001	MIPGAN-II	0.664	0.047	0.043	0.036	0.000	0.0009	0.0001	0.876	0.961$^{(13)}$
unibo-002	MIPGAN-II	0.011	0.016	0.028	0.083	0.000	0.0000	0.0000	0.751	0.962$^{(14)}$
ntnusub-000	MIPGAN-II	0.972	0.031	0.017	0.002	0.000	0.0000	0.0000	0.954	0.966$^{(15)}$

ntnucan-000	MIPGAN-II	0.994	0.016	0.004	0.001	0.000	0.0000	0.0000	0.959	0.966(16)
visteam-003	MIPGAN-II	0.008	0.844	0.916	0.653	0.000	0.0009	0.0001	0.929	0.972(17)
visteam-000	MIPGAN-II	0.511	0.438	0.290	0.459	0.000	0.0013	0.0001	0.929	0.974(18)
kinit-001	MIPGAN-II	0.810	0.016	0.008	0.060	0.000	0.0030	0.0005	0.972	0.996(19)
hdaarcface-001	MIPGAN-II	0.001	0.188	0.303	0.382	0.004	0.0041	0.0039	1.000	1.000(20)
hdabsif-004	MIPGAN-II	0.293	0.750	0.902	0.408	0.034	0.0870	0.0108	1.000	1.000(21)
hdadfr-002	MIPGAN-II	0.000	0.219	0.394	0.382	0.034	0.0871	0.0116	1.000	1.000(22)
hdadfr-003	MIPGAN-II	0.001	0.234	0.429	0.418	0.034	0.0980	0.0127	1.000	1.000(23)
hdadfr-006	MIPGAN-II	0.002	0.406	0.642	0.395	0.037	0.5145	0.0101	1.000	1.000(24)
hdafusion-001	MIPGAN-II	0.000	0.312	0.426	0.410	0.034	0.1026	0.0143	1.000	1.000(25)
hdalaplace-001	MIPGAN-II	0.228	0.734	0.905	0.715	0.034	0.0870	0.0108	1.000	1.000(26)
hdalbp-006	MIPGAN-II	0.002	0.969	0.969	0.791	0.034	0.1006	0.0142	1.000	1.000(27)
hdamag-001	MIPGAN-II	0.000	0.266	0.381	0.421	0.034	0.1024	0.0140	1.000	1.000(28)
hdawl-000	MIPGAN-II	0.032	0.984	0.994	0.864	0.758	0.9568	0.3556	1.000	1.000(29)
hdawl-002	MIPGAN-II	0.052	0.938	0.884	0.833	0.034	0.1141	0.0165	1.000	1.000(30)
secunet-001	MIPGAN-II	0.029	0.125	0.066	0.041	0.140	0.0479	0.0150	1.000	1.000(31)
unibo-001	MIPGAN-II	0.000	0.047	0.088	0.147	0.000	0.0000	0.0000	1.000	1.000(32)

4.3.3 Tier 3 - High Quality Morphs

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
secunet-004	Manual	0.023	0.106	0.031	0.093	0.000	0.0006	0.0003	0.180	0.449(1)
idemia-002	Manual	0.448	0.029	0.022	0.011	0.000	0.0003	0.0000	0.463	0.743(2)
idemia-003	Manual	0.442	0.051	0.009	0.025	0.000	0.0003	0.0000	0.699	0.743(3)
idemia-001	Manual	0.563	0.022	0.009	0.006	0.000	0.0003	0.0000	0.448	0.759(4)
secunet-003	Manual	0.062	0.057	0.026	0.061	0.000	0.0023	0.0006	0.281	0.848(5)
secunet-002	Manual	0.144	0.039	0.034	0.039	0.002	0.0059	0.0012	0.357	0.935(6)
visteam-003	Manual	0.029	0.884	0.916	0.653	0.000	0.0009	0.0001	0.872	0.937(7)
visteam-004	Manual	0.921	0.039	0.121	0.055	0.000	0.0009	0.0001	0.911	0.938(8)
visteam-002	Manual	0.845	0.043	0.068	0.054	0.000	0.0009	0.0001	0.954	0.979(9)
visteam-001	Manual	0.905	0.022	0.043	0.036	0.000	0.0009	0.0001	0.975	0.986(10)
wvudiff-001	Manual	0.458	0.527	0.654	0.417	0.005	0.0000	0.0000	0.989	0.990(11)
unibo-002	Manual	0.733	0.966	0.028	0.083	0.000	0.0000	0.0000	0.969	0.992(12)
ntnusub-000	Manual	0.995	0.004	0.017	0.002	0.000	0.0000	0.0000	0.967	0.994(13)
visteamicao-000	Manual	0.987	0.003	0.028	0.005	0.000	0.0000	0.0000	0.981	0.994(14)
kinit-001	Manual	0.967	0.026	0.008	0.060	0.003	0.0030	0.0005	0.989	0.994(15)
visteam-000	Manual	0.764	0.142	0.290	0.459	0.000	0.0013	0.0001	0.981	0.997(16)
ntnucan-000	Manual	1.000	0.003	0.004	0.001	0.000	0.0000	0.0000	0.992	0.997(17)
visionbox-000	Manual	0.136	0.194	0.162	0.102	0.000	0.0008	0.0001	0.997	0.997(18)
hdaarcface-001	Manual	0.010	0.417	0.303	0.382	0.005	0.0041	0.0039	1.000	1.000(19)
hdabsif-004	Manual	0.491	0.500	0.902	0.408	0.003	0.0870	0.0108	1.000	1.000(20)
hdadfr-002	Manual	0.010	0.382	0.394	0.382	0.000	0.0871	0.0116	1.000	1.000(21)
hdadfr-003	Manual	0.008	0.398	0.429	0.418	0.000	0.0980	0.0127	1.000	1.000(22)
hdadfr-006	Manual	0.209	0.378	0.642	0.395	0.029	0.5145	0.0101	1.000	1.000(23)
hdafusion-001	Manual	0.005	0.388	0.426	0.410	0.000	0.1026	0.0143	1.000	1.000(24)

hdalaplace-001	Manual	0.135	0.713	0.905	0.715	0.000	0.0870	0.0108	1.000	1.000[25]
hdalbp-006	Manual	0.143	0.801	0.969	0.791	0.000	0.1006	0.0142	1.000	1.000[26]
hdamag-001	Manual	0.002	0.441	0.381	0.421	0.000	0.1024	0.0140	1.000	1.000[27]
hdawl-000	Manual	0.036	0.898	0.994	0.864	0.649	0.9568	0.3556	1.000	1.000[28]
hdawl-002	Manual	0.161	0.758	0.884	0.833	0.000	0.1141	0.0165	1.000	1.000[29]
secunet-001	Manual	0.154	0.090	0.066	0.041	0.054	0.0479	0.0150	1.000	1.000[30]
unibo-001	Manual	0.610	0.989	0.088	0.147	0.000	0.0000	0.0000	1.000	1.000[31]
visionbox-001	Manual	0.918	0.144	0.188	0.047	0.000	0.0008	0.0002	1.000	1.000[32]

Algorithm	Dataset	MACER*	$BSCER_q$ **	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_q$	MACER @ $BSCER_q$=0.01	MACER @ $BSCER_q$=0.003
idemia-003	Print + Scanned	0.055	0.016	0.000	0.000	0.125	0.179[1]
secunet-004	Print + Scanned	0.019	0.039	0.000	0.000	0.073	0.263[2]
secunet-003	Print + Scanned	0.050	0.023	0.001	0.000	0.105	0.433[3]
idemia-001	Print + Scanned	0.349	0.005	0.000	0.000	0.293	0.440[4]
idemia-002	Print + Scanned	0.103	0.020	0.000	0.000	0.182	0.455[5]
unibo-002	Print + Scanned	0.143	0.023	0.000	0.000	0.280	0.647[6]
visteamicao-000	Print + Scanned	0.816	0.006	0.000	0.000	0.751	0.859[7]
secunet-002	Print + Scanned	0.047	0.036	0.004	0.002	0.176	0.919[8]
visionbox-001	Print + Scanned	0.115	0.471	0.001	0.000	0.874	0.923[9]
visionbox-000	Print + Scanned	0.013	0.439	0.001	0.000	0.905	0.929[10]
visteam-004	Print + Scanned	0.886	0.091	0.001	0.000	0.915	0.947[11]
visteam-003	Print + Scanned	0.063	0.813	0.001	0.000	0.926	0.978[12]
visteam-002	Print + Scanned	0.786	0.037	0.001	0.000	0.917	0.981[13]
wvudiff-001	Print + Scanned	0.072	0.878	0.000	0.000	0.953	0.985[14]
visteam-001	Print + Scanned	0.846	0.026	0.001	0.000	0.930	0.986[15]
kinit-001	Print + Scanned	0.994	0.002	0.002	0.001	0.974	0.990[16]
ntnucan-000	Print + Scanned	0.995	0.004	0.000	0.000	0.990	0.996[17]
ntnusub-000	Print + Scanned	0.985	0.016	0.000	0.000	0.991	0.997[18]
visteam-000	Print + Scanned	0.657	0.292	0.001	0.000	0.987	0.998[19]
hdaarcface-001	Print + Scanned	0.007	0.376	0.003	0.001	1.000	1.000[20]
hdabsif-004	Print + Scanned	0.004	0.996	0.067	0.065	1.000	1.000[21]
hdadfr-002	Print + Scanned	0.006	0.435	0.067	0.065	1.000	1.000[22]
hdadfr-003	Print + Scanned	0.006	0.464	0.076	0.075	1.000	1.000[23]
hdadfr-006	Print + Scanned	0.037	0.630	0.488	0.479	1.000	1.000[24]
hdafusion-001	Print + Scanned	0.001	0.578	0.081	0.081	1.000	1.000[25]
hdalaplace-001	Print + Scanned	0.020	0.957	0.067	0.065	1.000	1.000[26]
hdalbp-006	Print + Scanned	0.017	0.917	0.080	0.082	1.000	1.000[27]
hdamag-001	Print + Scanned	0.005	0.399	0.079	0.080	1.000	1.000[28]
hdawl-000	Print + Scanned	0.006	0.989	0.950	0.948	1.000	1.000[29]
hdawl-002	Print + Scanned	0.113	0.833	0.090	0.091	1.000	1.000[30]
secunet-001	Print + Scanned	0.063	0.059	0.037	0.037	1.000	1.000[31]
unibo-001	Print + Scanned	0.084	0.054	0.000	0.000	1.000	1.000[32]

4.4 Two-image Differential Morph Detection with Subject Metadata

Two face photos are provided to the algorithm: the first being a suspected morph and the second probe image representing a known, non-morphed face image of one of the subjects contributing to the morph (e.g., live capture image from an eGate). In the case that the first image is a bona fide photo, then the second image will be a known non-morphed image of the same subject taken on a different day. **Additionally,** information about the subject (sex, age of the subject at the time the probe image is taken, and the age/time difference between the suspected morph and the live probe image) is provided as input to the algorithm. Operationally, this information might be derived from data read from the machine readable zone of a passport for example.

4.4.1 Tier 1 - Low Quality Morphs

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
visionbox-001	Global Morph	0.006	0.087	0.188	0.047	0.000	0.0008	0.0002	0.013	$0.014^{(1)}$
visionbox-000	Global Morph	0.004	0.063	0.162	0.102	0.000	0.0008	0.0001	0.017	$0.021^{(2)}$
secunet-004	Global Morph	0.064	0.000	0.025	0.034	0.000	0.0006	0.0003	0.156	$0.368^{(3)}$
secunet-003	Global Morph	0.038	0.000	0.044	0.122	0.000	0.0023	0.0006	0.289	$0.773^{(4)}$
visteam-003	Global Morph	0.003	0.819	0.916	0.653	0.000	0.0009	0.0001	0.538	$0.867^{(5)}$
visteam-002	Global Morph	0.460	0.031	0.068	0.054	0.000	0.0009	0.0001	0.747	$0.879^{(6)}$
visteam-004	Global Morph	0.207	0.220	0.121	0.055	0.000	0.0009	0.0001	0.634	$0.889^{(7)}$
secunet-002	Global Morph	0.112	0.000	0.034	0.029	0.000	0.0059	0.0012	0.268	$0.892^{(8)}$
visteamicao-000	Global Morph	0.847	0.000	0.028	0.005	0.000	0.0000	0.0000	0.767	$0.897^{(9)}$
visteam-001	Global Morph	0.638	0.008	0.043	0.036	0.000	0.0009	0.0001	0.858	$0.947^{(10)}$
secunet-001	Global Morph	0.144	0.000	0.063	0.031	0.000	0.0479	0.0150	1.000	$1.000^{(11)}$

4.4.2 Tier 2 - Automated Morphs

Algorithm	Dataset	MACER*	$BSCER_q$**	$BSCER_v$** (visa)	$BSCER_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process $(Bona\ Fides)_v$	Failure to Process $(Bona\ Fides)_m$	MACER @ $BSCER_m$ =0.01	MACER @ $BSCER_m$ =0.003
visionbox-001	Local Morph	0.003	0.087	0.188	0.047	0.000	0.0008	0.0002	0.006	$0.008^{(1)}$
visionbox-000	Local Morph	0.001	0.063	0.162	0.102	0.000	0.0008	0.0001	0.006	$0.009^{(2)}$
secunet-004	Local Morph	0.042	0.000	0.025	0.034	0.000	0.0006	0.0003	0.111	$0.319^{(3)}$
secunet-003	Local Morph	0.024	0.000	0.044	0.122	0.000	0.0023	0.0006	0.260	$0.764^{(4)}$
visteam-003	Local Morph	0.001	0.819	0.916	0.653	0.000	0.0009	0.0001	0.450	$0.782^{(5)}$

* MACER: This is the rate that morphs that are not detected. Lower values are better.
** BSCER: This is the rate that bona fides that were mistaken for morphs. Lower values are better.
For each dataset, the entries are ordered by the metric in the last table column.
Entries with - in them mean results are missing either due to the algorithm not being able to process the entire dataset OR results are still currently being generated.

visteamicao-000	Local Morph	0.727	0.000	0.028	0.005	0.000	0.0000	0.0000	0.638	0.809$^{(6)}$
visteam-002	Local Morph	0.336	0.031	0.068	0.054	0.000	0.0009	0.0001	0.662	0.814$^{(7)}$
visteam-004	Local Morph	0.165	0.220	0.121	0.055	0.000	0.0009	0.0001	0.612	0.883$^{(8)}$
secunet-002	Local Morph	0.090	0.000	0.034	0.029	0.000	0.0059	0.0012	0.237	0.889$^{(9)}$
visteam-001	Local Morph	0.553	0.008	0.043	0.036	0.000	0.0009	0.0001	0.789	0.912$^{(10)}$
secunet-001	Local Morph	0.115	0.000	0.063	0.031	0.000	0.0479	0.0150	1.000	1.000$^{(11)}$

Algorithm	Dataset	MACER*	BSCER$_q$**	BSCER$_v$** (visa)	BSCER$_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process (Bona Fides)$_v$	Failure to Process (Bona Fides)$_m$	MACER @ BSCER$_m$ =0.01	MACER @ BSCER$_m$ =0.003
visionbox-001	Local Morph Colorized Average	0.003	0.087	0.188	0.047	0.000	0.0008	0.0002	0.009	0.011$^{(1)}$
visionbox-000	Local Morph Colorized Average	0.001	0.063	0.162	0.102	0.000	0.0008	0.0001	0.013	0.020$^{(2)}$
secunet-004	Local Morph Colorized Average	0.048	0.000	0.025	0.034	0.000	0.0006	0.0003	0.131	0.337$^{(3)}$
secunet-003	Local Morph Colorized Average	0.030	0.000	0.044	0.122	0.000	0.0023	0.0006	0.280	0.786$^{(4)}$
visteam-003	Local Morph Colorized Average	0.003	0.819	0.916	0.653	0.000	0.0009	0.0001	0.464	0.799$^{(5)}$
visteamicao-000	Local Morph Colorized Average	0.793	0.000	0.028	0.005	0.000	0.0000	0.0000	0.708	0.858$^{(6)}$
visteam-002	Local Morph Colorized Average	0.391	0.031	0.068	0.054	0.000	0.0009	0.0001	0.724	0.868$^{(7)}$
visteam-004	Local Morph Colorized Average	0.267	0.220	0.121	0.055	0.000	0.0009	0.0001	0.663	0.900$^{(8)}$
secunet-002	Local Morph Colorized Average	0.101	0.000	0.034	0.029	0.000	0.0059	0.0012	0.249	0.904$^{(9)}$
visteam-001	Local Morph Colorized Average	0.593	0.008	0.043	0.036	0.000	0.0009	0.0001	0.839	0.938$^{(10)}$
secunet-001	Local Morph Colorized Average	0.133	0.000	0.063	0.031	0.000	0.0479	0.0150	1.000	1.000$^{(11)}$

Algorithm	Dataset	MACER*	BSCER$_q$**	BSCER$_v$** (visa)	BSCER$_m$** (mugshot)	Failure to Process (Morphs)	Failure to Process (Bona Fides)$_v$	Failure to Process (Bona Fides)$_m$	MACER @ BSCER$_m$ =0.01	MACER @ BSCER$_m$ =0.003
visionbox-001	Local Morph Colorized Match	0.006	0.087	0.188	0.047	0.000	0.0008	0.0002	0.014	0.022$^{(1)}$
visionbox-000	Local Morph Colorized Match	0.008	0.063	0.162	0.102	0.000	0.0008	0.0001	0.018	0.028$^{(2)}$

secunet-004	Local Morph Colorized Match	0.053	0.000	0.025	0.034	0.000	0.0006	0.0003	0.131	$0.340^{(3)}$
secunet-003	Local Morph Colorized Match	0.031	0.000	0.044	0.122	0.000	0.0023	0.0006	0.287	$0.790^{(4)}$
visteam-003	Local Morph Colorized Match	0.003	0.819	0.916	0.653	0.000	0.0009	0.0001	0.488	$0.808^{(5)}$
visteam-004	Local Morph Colorized Match	0.328	0.220	0.121	0.055	0.000	0.0009	0.0001	0.710	$0.906^{(6)}$
secunet-002	Local Morph Colorized Match	0.108	0.000	0.034	0.029	0.000	0.0059	0.0012	0.259	$0.911^{(7)}$
visteamicao-000	Local Morph Colorized Match	0.868	0.000	0.028	0.005	0.000	0.0000	0.0000	0.811	$0.916^{(8)}$
visteam-002	Local Morph Colorized Match	0.545	0.031	0.068	0.054	0.000	0.0009	0.0001	0.816	$0.918^{(9)}$
visteam-001	Local Morph Colorized Match	0.694	0.008	0.043	0.036	0.000	0.0009	0.0001	0.878	$0.957^{(10)}$
secunet-001	Local Morph Colorized Match	0.140	0.000	0.063	0.031	0.000	0.0479	0.0150	1.000	$1.000^{(11)}$

Algorithm	Dataset	MACER*	BSCER_q**	BSCER_v** (visa)	BSCER_m** (mugshot)	Failure to Process (Morphs)	Failure to Process $(\text{Bona Fides})_v$	Failure to Process $(\text{Bona Fides})_m$	MACER @ BSCER_m =0.01	MACER @ BSCER_m =0.003
visionbox-001	Visa-Border	0.002	0.138	0.188	0.047	0.0002	0.0008	0.0002	0.003	$0.004^{(1)}$
visionbox-000	Visa-Border	0.001	0.115	0.162	0.102	0.0002	0.0008	0.0001	0.004	$0.005^{(2)}$
secunet-004	Visa-Border	0.021	0.013	0.025	0.034	0.0002	0.0006	0.0003	0.077	$0.263^{(3)}$
visteamicao-000	Visa-Border	0.460	0.023	0.028	0.005	0.0000	0.0000	0.0000	0.388	$0.535^{(4)}$
secunet-003	Visa-Border	0.016	0.027	0.044	0.122	0.0007	0.0023	0.0006	0.268	$0.821^{(5)}$
visteam-003	Visa-Border	0.011	0.921	0.916	0.653	0.0002	0.0009	0.0001	0.453	$0.841^{(6)}$
secunet-002	Visa-Border	0.047	0.019	0.034	0.029	0.0027	0.0059	0.0012	0.161	$0.903^{(7)}$
visteam-002	Visa-Border	0.513	0.053	0.068	0.054	0.0002	0.0009	0.0001	0.815	$0.928^{(8)}$
visteam-001	Visa-Border	0.659	0.032	0.043	0.036	0.0002	0.0009	0.0001	0.844	$0.936^{(9)}$
visteam-004	Visa-Border	0.556	0.116	0.121	0.055	0.0002	0.0009	0.0001	0.882	$0.959^{(10)}$
secunet-001	Visa-Border	0.058	0.043	0.063	0.031	0.0343	0.0479	0.0150	1.000	$1.000^{(11)}$

4.4.3 Tier 3 - High Quality Morphs

Algorithm	Dataset	MACER*	BSCER_q**	Failure to Process (Morphs)	Failure to Process $(\text{Bona Fides})_q$	MACER @ BSCER_q=0.01	MACER @ BSCER_q=0.003
secunet-004	Print + Scanned	0.020	0.031	0.000	0.000	0.065	$0.175^{(1)}$
secunet-003	Print + Scanned	0.024	0.044	0.001	0.000	0.102	$0.471^{(2)}$
visteamicao-000	Print + Scanned	0.816	0.006	0.000	0.000	0.751	$0.859^{(3)}$
visionbox-001	Print + Scanned	0.115	0.471	0.001	0.000	0.874	$0.923^{(4)}$

visionbox-000	Print + Scanned	0.013	0.439	0.001	0.000	0.905	$0.929^{(5)}$
secunet-002	Print + Scanned	0.045	0.035	0.004	0.002	0.169	$0.940^{(6)}$
visteam-003	Print + Scanned	0.063	0.813	0.001	0.000	0.568	$0.977^{(7)}$
visteam-002	Print + Scanned	0.786	0.037	0.001	0.000	0.917	$0.981^{(8)}$
visteam-001	Print + Scanned	0.846	0.026	0.001	0.000	0.930	$0.986^{(9)}$
visteam-004	Print + Scanned	0.886	0.091	0.001	0.000	0.980	$0.988^{(10)}$
secunet-001	Print + Scanned	0.058	0.057	0.037	0.037	1.000	$1.000^{(11)}$

4.5 1:1 Comparison (Morph-resistant Face Recognition)

Two face samples are provided to the algorithm for one-to-one comparison of whether the two images are of the same subject. The expected behavior from the algorithm is to be able to correctly reject comparisons of morphed images against constituents that contributed to the morph, while correctly accepting/matching comparisons of unaltered/non-morphed images of the same subject and rejecting comparisons of different people.

4.5.1 Tier 2 - Automated Morphs

Algorithm	Dataset	FNMR* @ FMR=0.001	MMPMR** @ FMR=0.001	RMMR*** @ FMR=0.001	FNMR @ FMR=0.0001	MMPMR @ FMR=0.0001	RMMR @ FMR=0.0001
visteam-001	Visa-Border	0.040	0.230	0.270	0.112	0.048	0.160(1)
visteam-002	Visa-Border	0.062	0.185	0.247	0.154	0.058	0.212(2)
visteam-004	Visa-Border	0.011	0.407	0.418	0.021	0.212	0.233(3)
visteamicao-000	Visa-Border	0.014	0.405	0.420	0.027	0.234	0.261(4)
visteam-003	Visa-Border	0.007	0.625	0.632	0.011	0.313	0.323(5)
visionbox-000	Visa-Border	0.004	0.990	0.994	0.005	0.577	0.582(6)
visionbox-001	Visa-Border	0.004	0.989	0.993	0.005	0.627	0.632(7)

* FNMR: This is the false non-match rate on regular non-morphed photos. Lower values are better.
** MMPMR: This is the rate that both subjects erroneously authenticated against a morphed photo. Lower values are better.
*** RMMR: This is the relative morph match rate, which assesses MMPMR (morph vulnerability) relative to FNMR (general algorithm accuracy). Lower values are better.
For each dataset, the entries are ordered by the metric in the last table column.
Entries with - in them mean results are missing either due to the algorithm not being able to process the entire dataset OR results are still currently being generated.

4.6 DET Analyses

4.6.1 Tier 1 - Low Quality Morphs

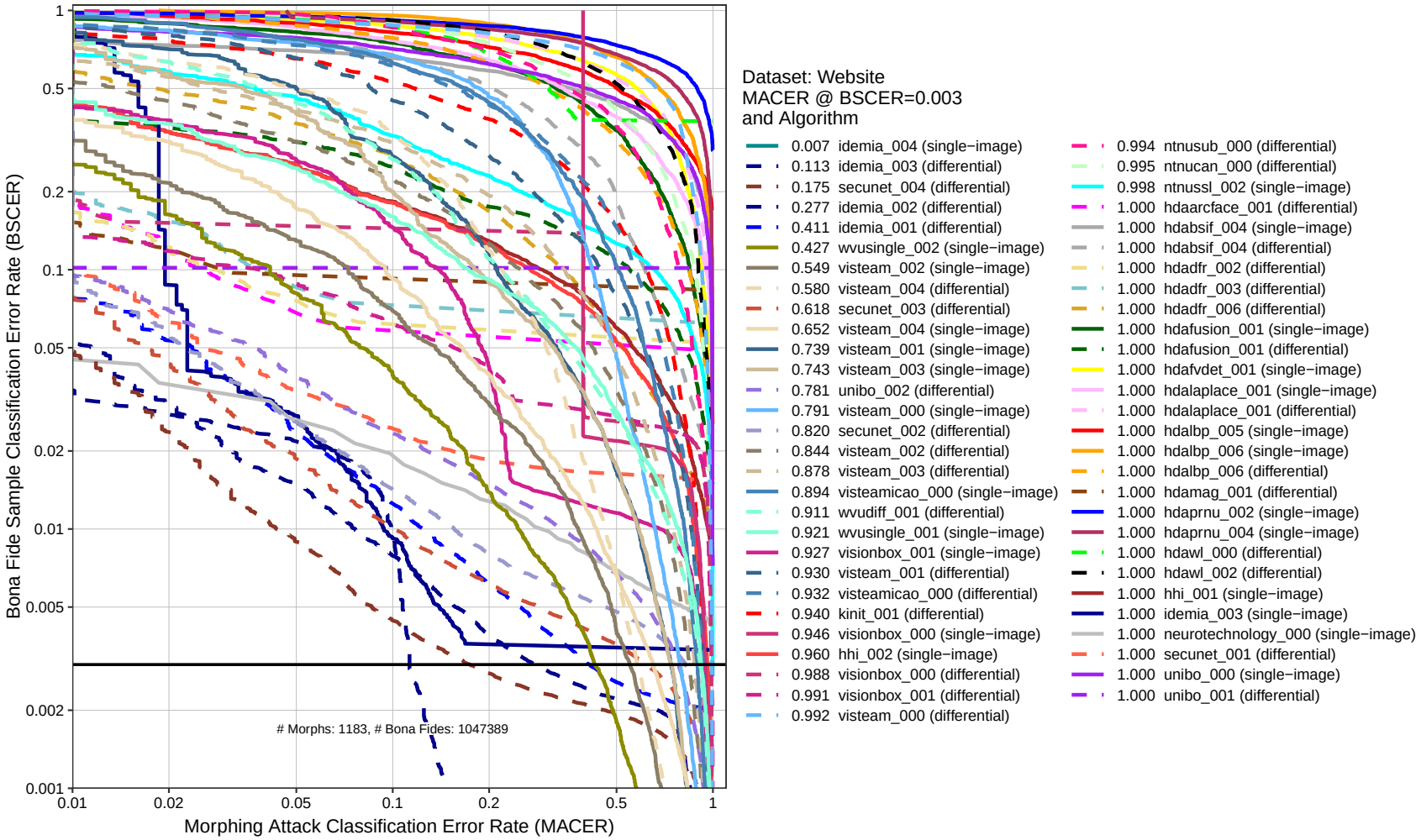


Figure 4: DET plot. This chart plots BSCER as a function of MACER. The x-axis is the rate morphs are not detected and the y-axis is the rate that bona fide images are falsely classified as morphs. The horizontal black line represents BSCER=0.003.

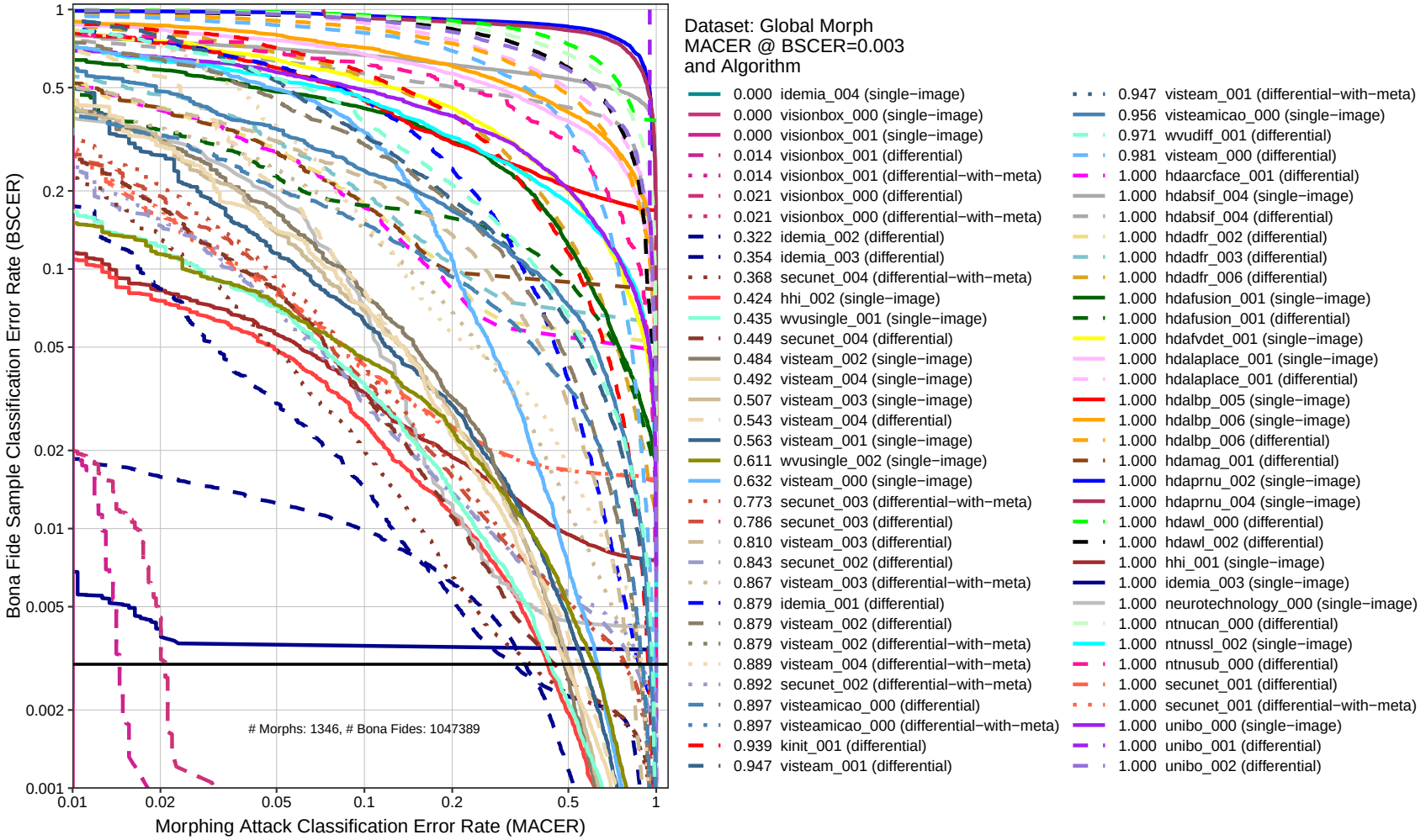


Figure 5: DET plot. This chart plots BSCER as a function of MACER. The x-axis is the rate morphs are not detected and the y-axis is the rate that bona fide images are falsely classified as morphs. The horizontal black line represents BSCER=0.003.

4.6.2 Tier 2 - Automated Morphs

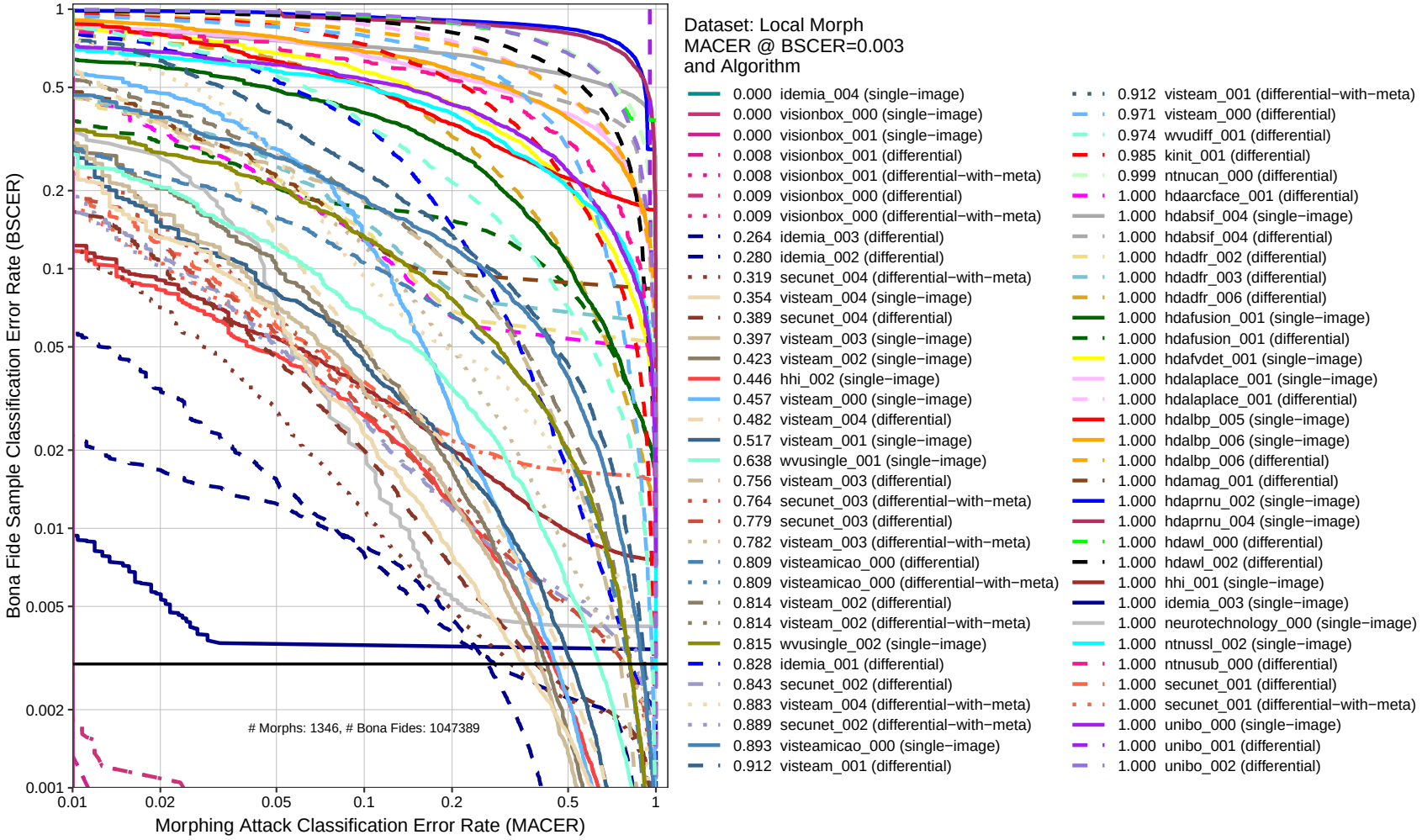


Figure 6: DET plot. This chart plots BSCER as a function of MACER. The x-axis is the rate morphs are not detected and the y-axis is the rate that bona fide images are falsely classified as morphs. The horizontal black line represents BSCER=0.003.

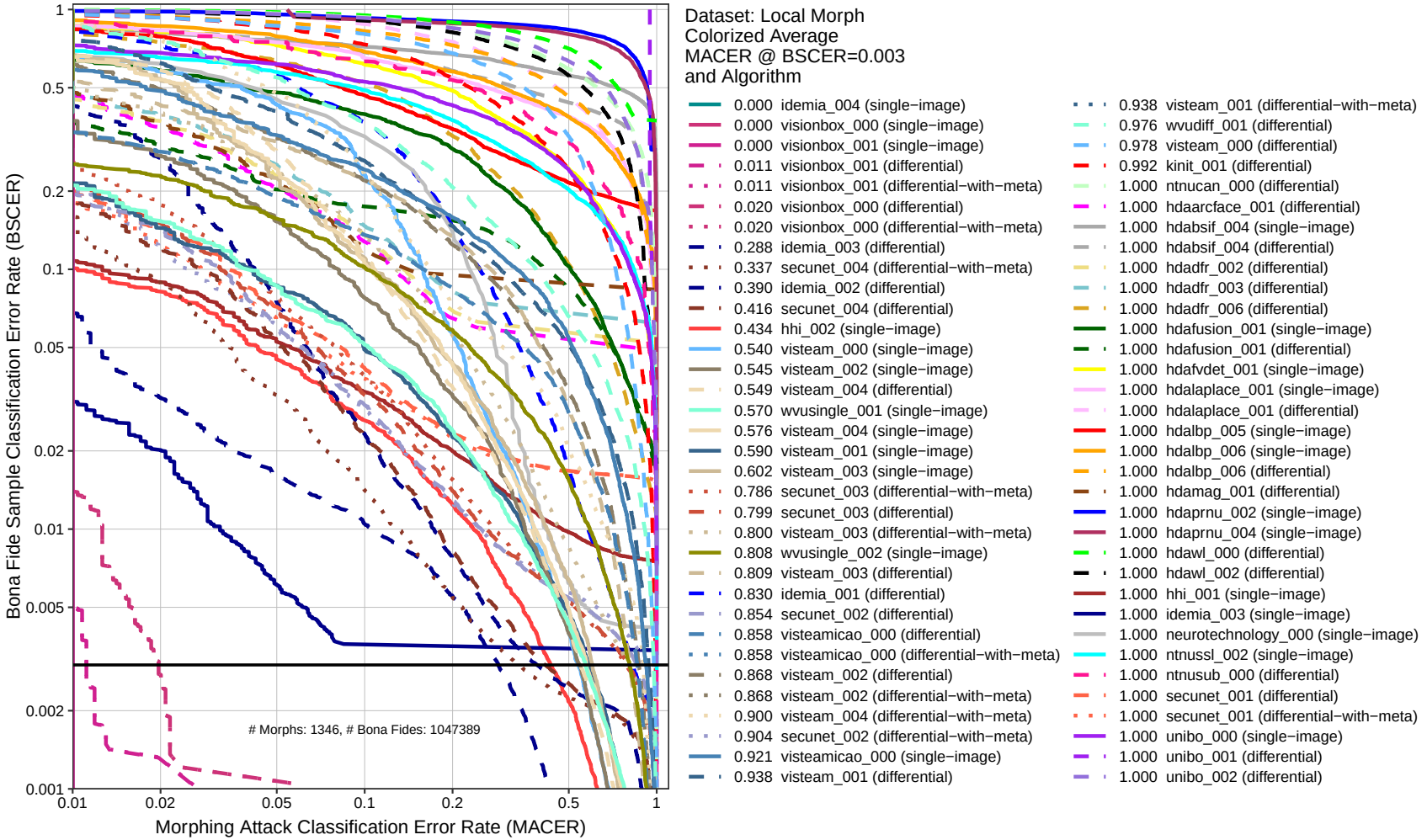


Figure 7: DET plot. This chart plots BSCER as a function of MACER. The x-axis is the rate morphs are not detected and the y-axis is the rate that bona fide images are falsely classified as morphs. The horizontal black line represents BSCER=0.003.

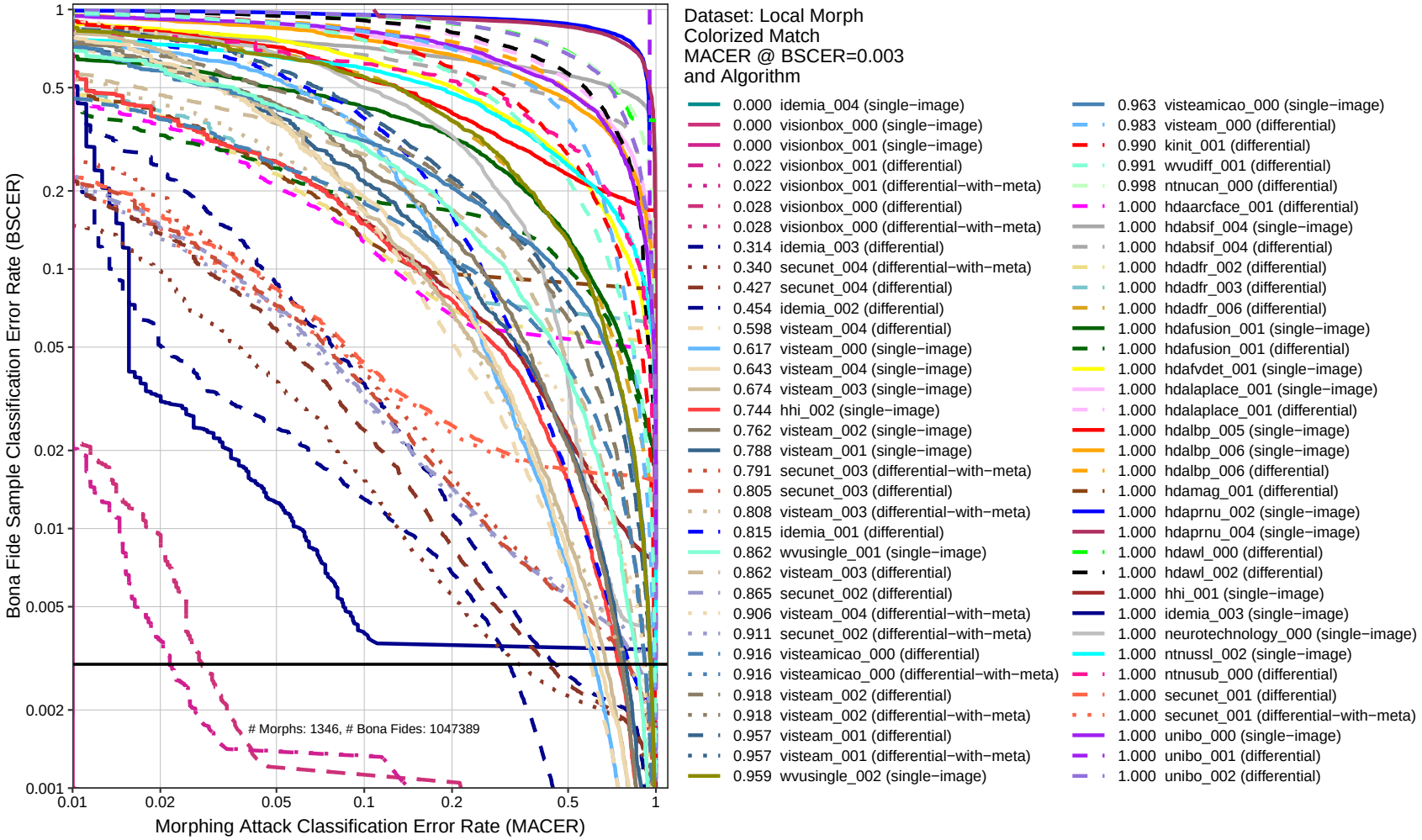


Figure 8: DET plot. This chart plots BSCER as a function of MACER. The x-axis is the rate morphs are not detected and the y-axis is the rate that bona fide images are falsely classified as morphs. The horizontal black line represents BSCER=0.003.

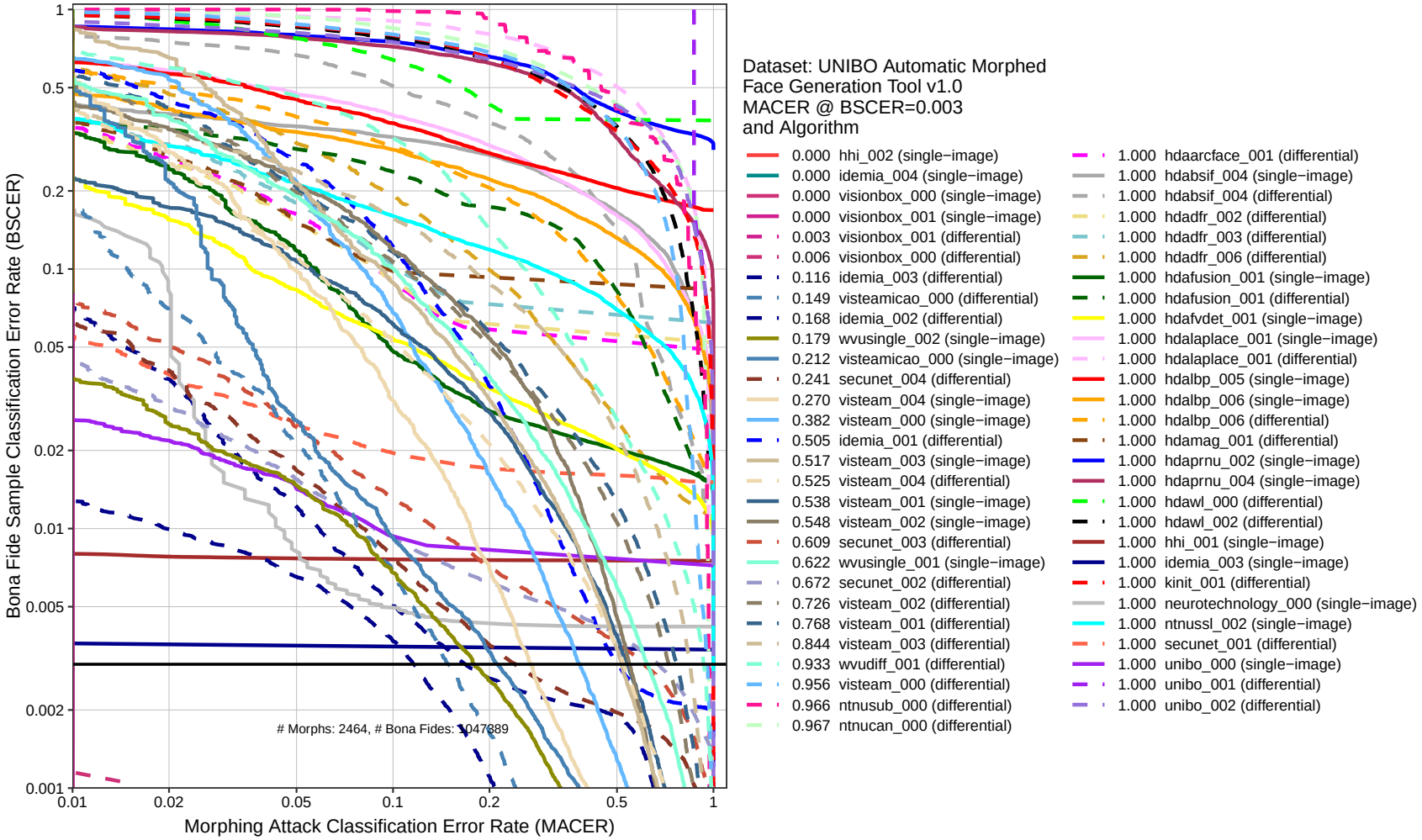


Figure 9: DET plot. This chart plots BSCER as a function of MACER. The x-axis is the rate morphs are not detected and the y-axis is the rate that bona fide images are falsely classified as morphs. The horizontal black line represents BSCER=0.003.

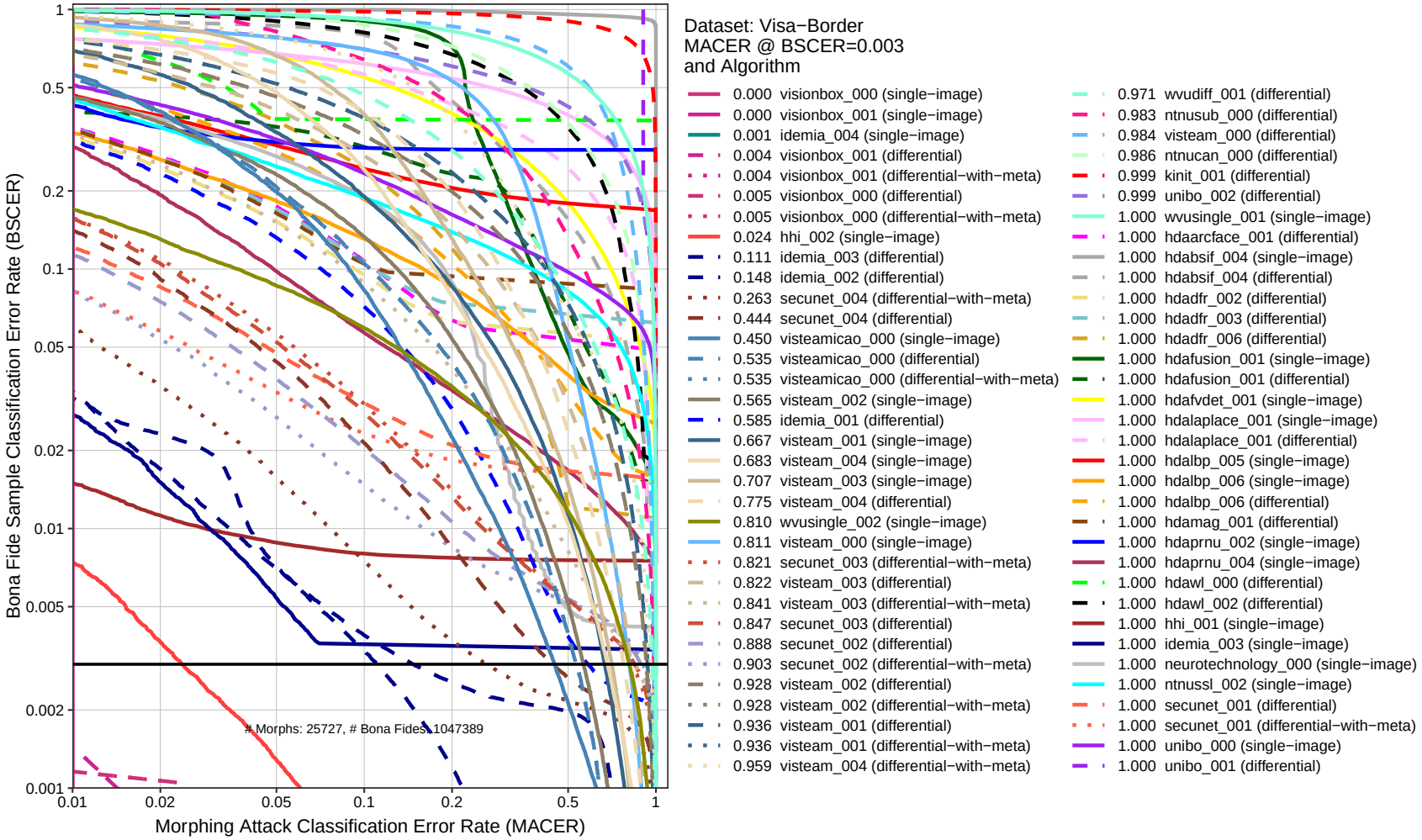


Figure 10: DET plot. This chart plots BSCER as a function of MACER. The x-axis is the rate morphs are not detected and the y-axis is the rate that bona fide images are falsely classified as morphs. The horizontal black line represents BSCER=0.003.

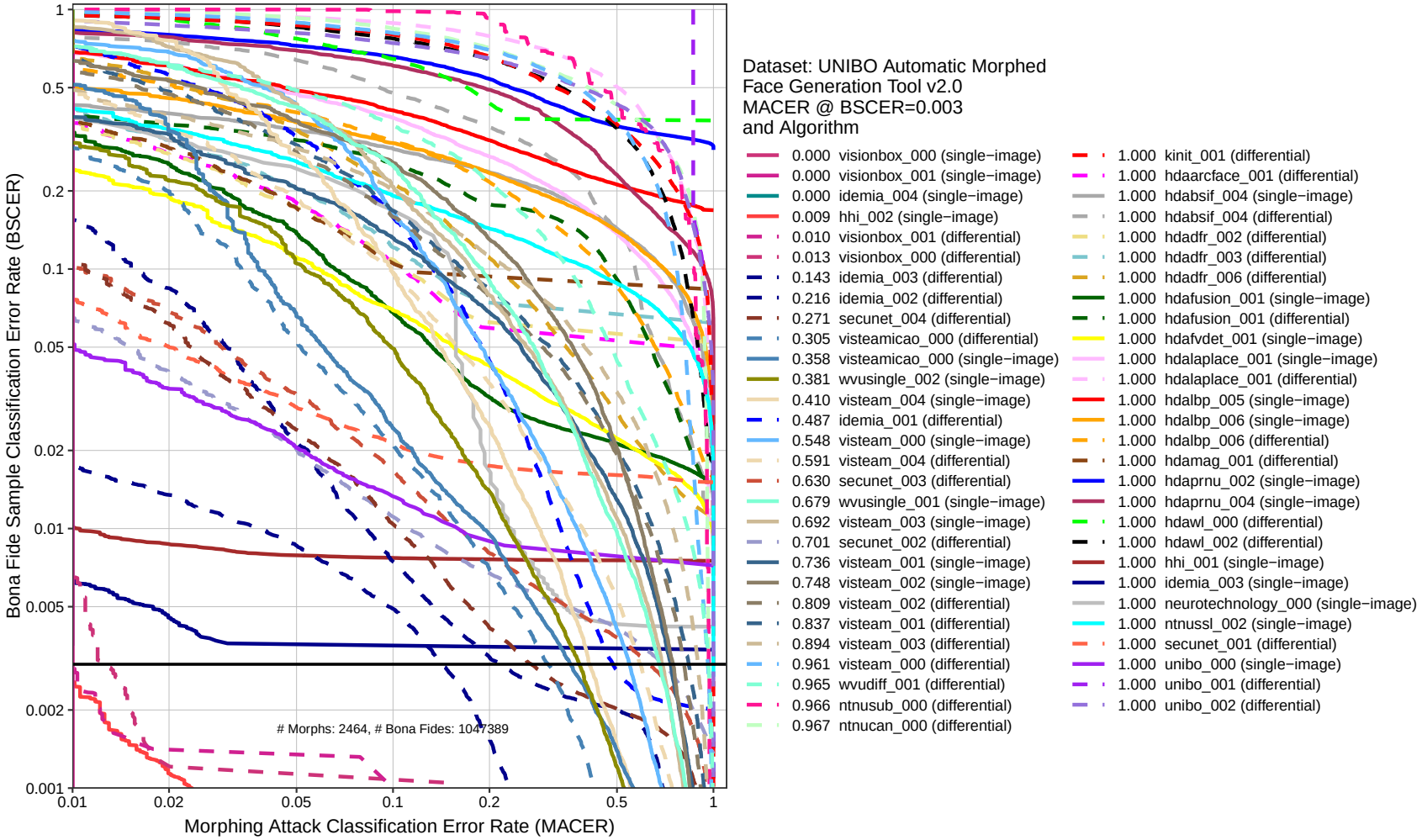


Figure 11: DET plot. This chart plots BSCER as a function of MACER. The x-axis is the rate morphs are not detected and the y-axis is the rate that bona fide images are falsely classified as morphs. The horizontal black line represents BSCER=0.003.

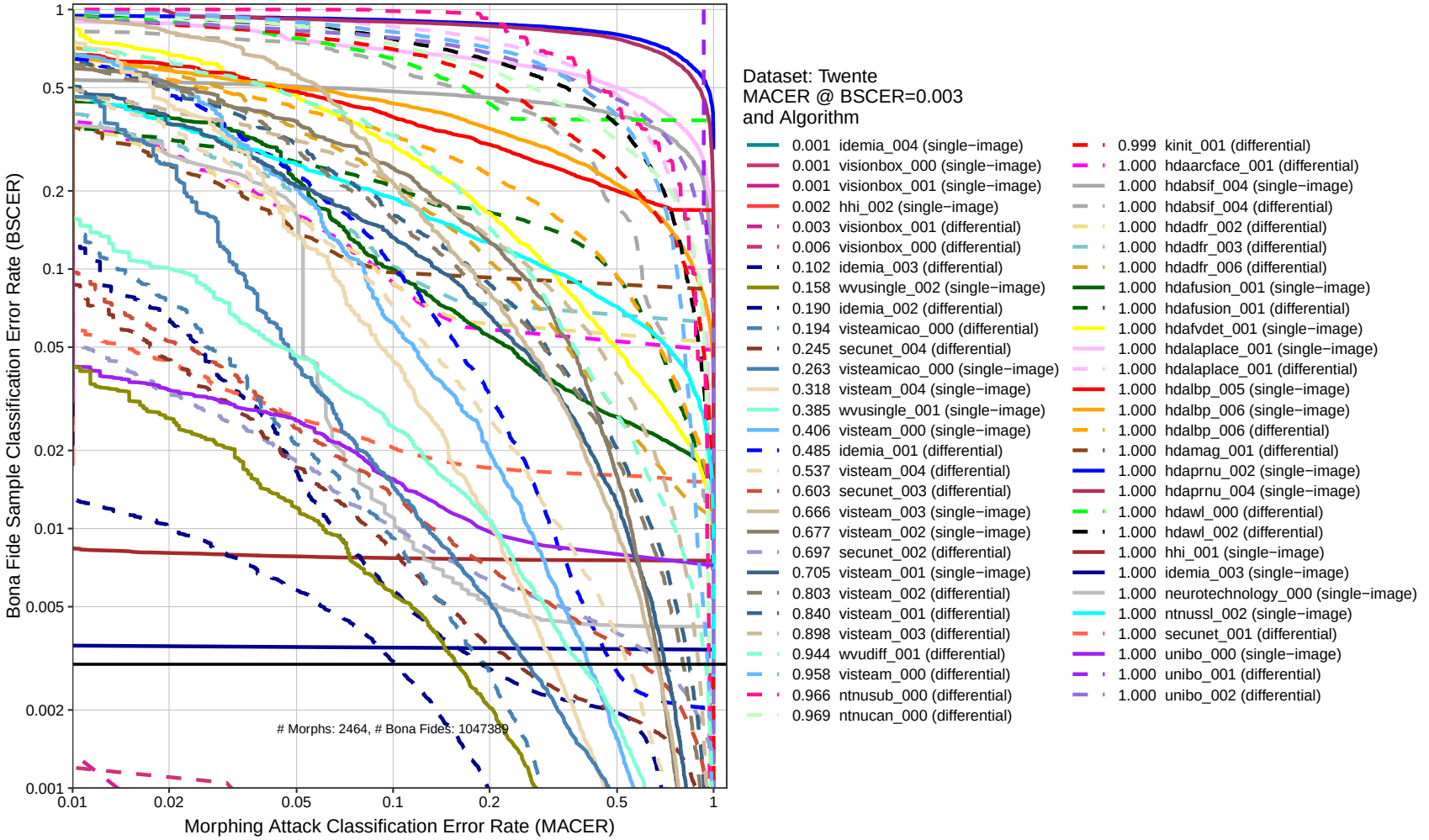


Figure 12: DET plot. This chart plots BSCER as a function of MACER. The x-axis is the rate morphs are not detected and the y-axis is the rate that bona fide images are falsely classified as morphs. The horizontal black line represents BSCER=0.003.

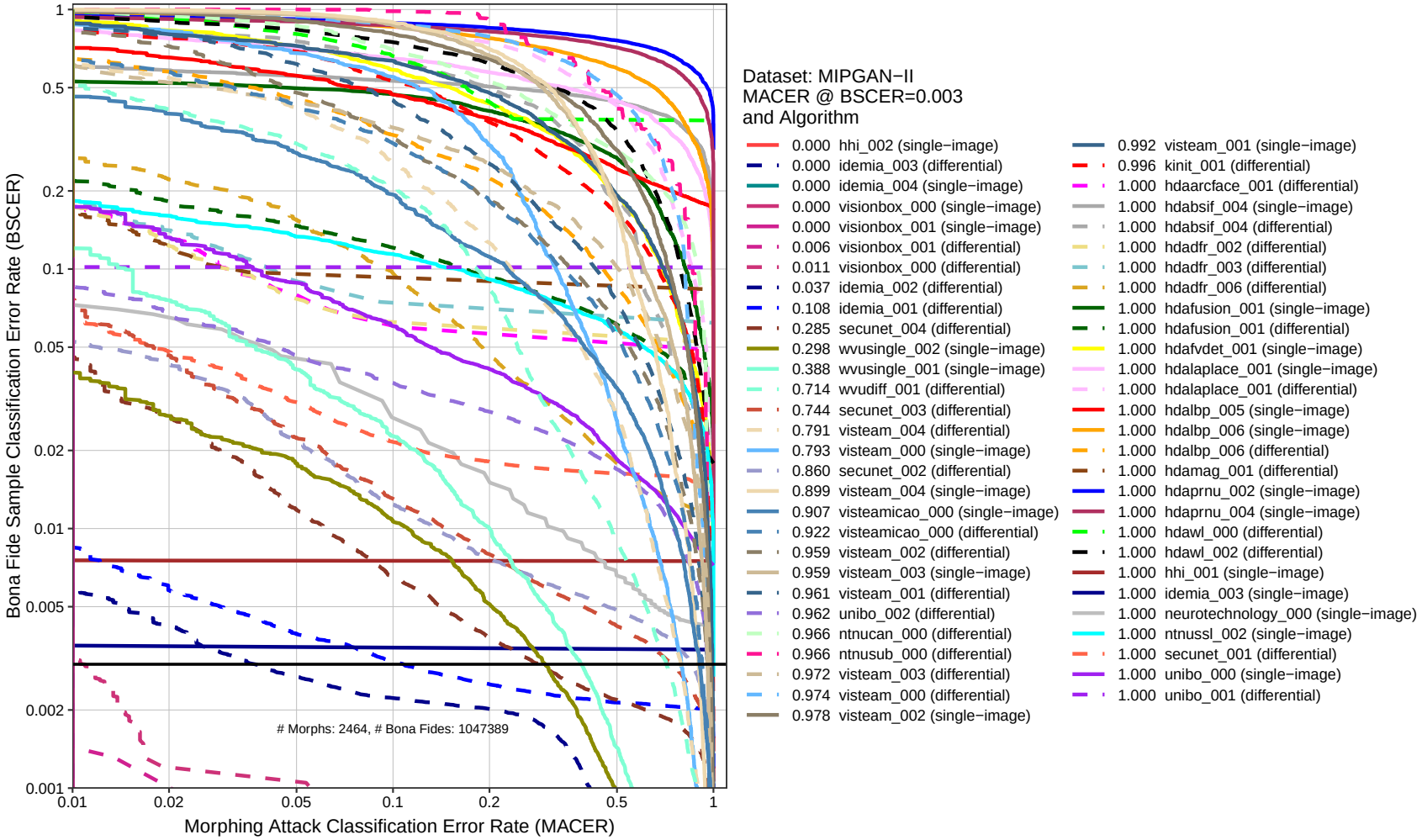


Figure 13: DET plot. This chart plots BSCER as a function of MACER. The x-axis is the rate morphs are not detected and the y-axis is the rate that bona fide images are falsely classified as morphs. The horizontal black line represents BSCER=0.003.

4.6.3 Tier 3 - High Quality Morphs

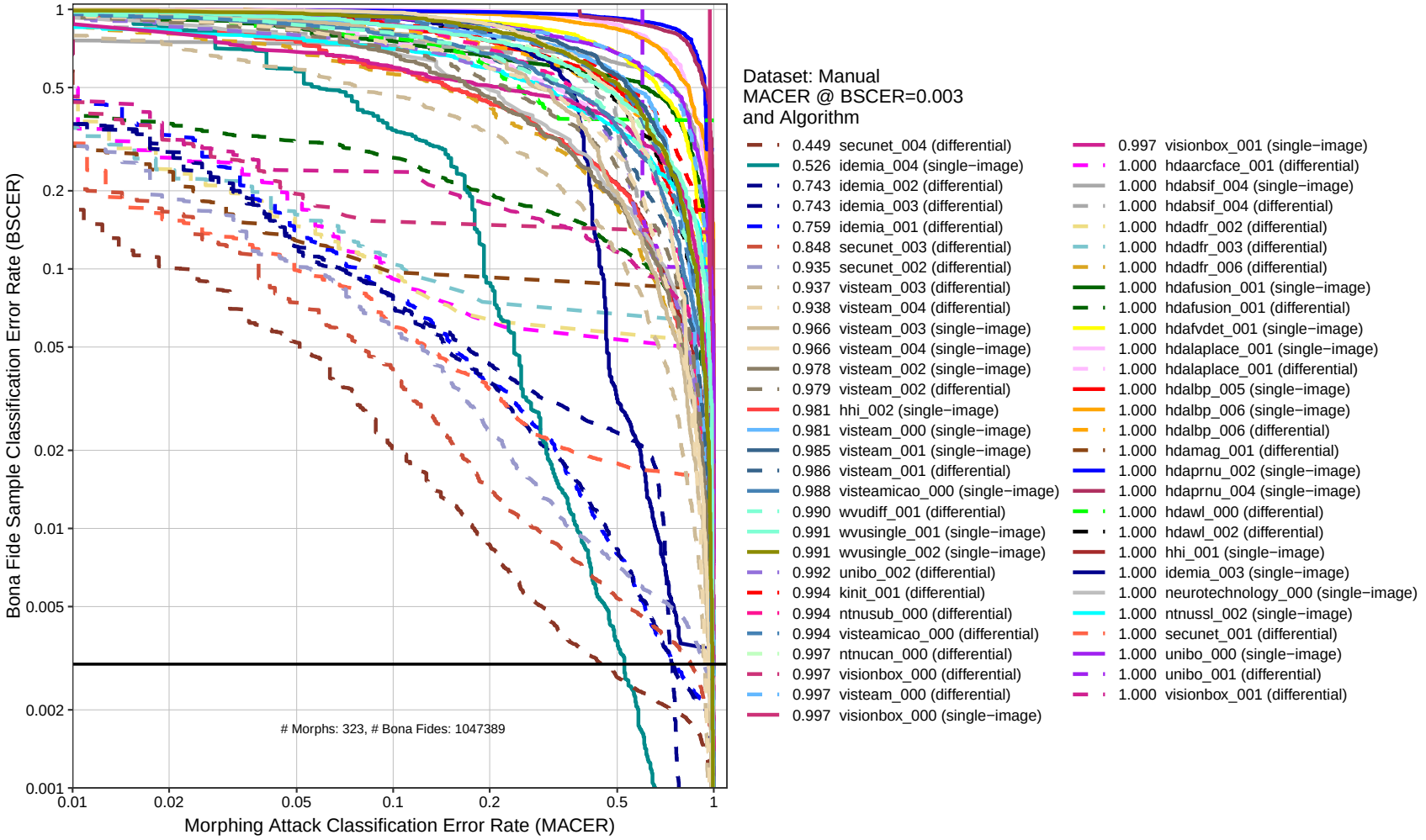


Figure 14: DET plot. This chart plots BSCER as a function of MACER. The x-axis is the rate morphs are not detected and the y-axis is the rate that bona fide images are falsely classified as morphs. The horizontal black line represents BSCER=0.003.

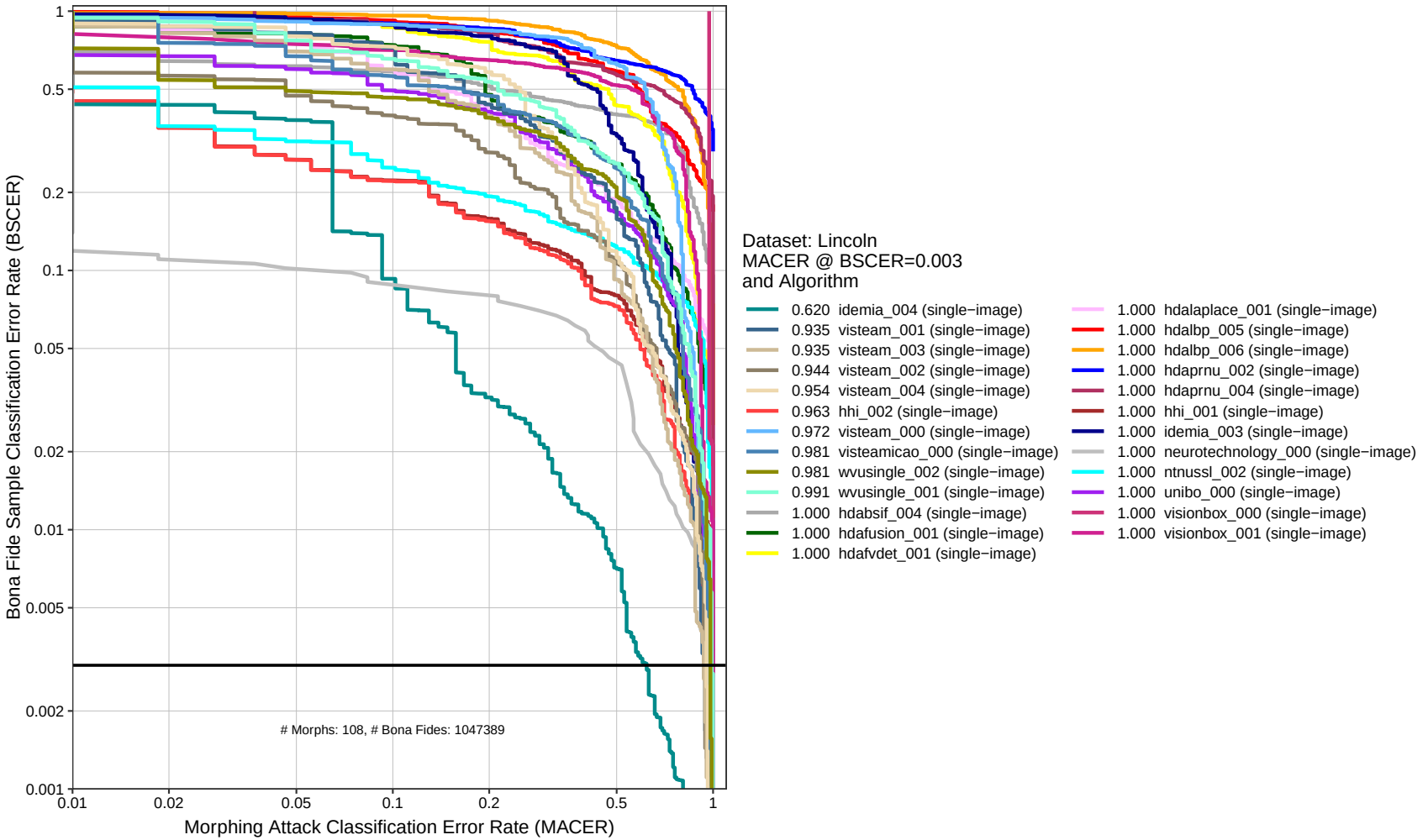


Figure 15: DET plot. This chart plots BSCER as a function of MACER. The x-axis is the rate morphs are not detected and the y-axis is the rate that bona fide images are falsely classified as morphs. The horizontal black line represents BSCER=0.003.

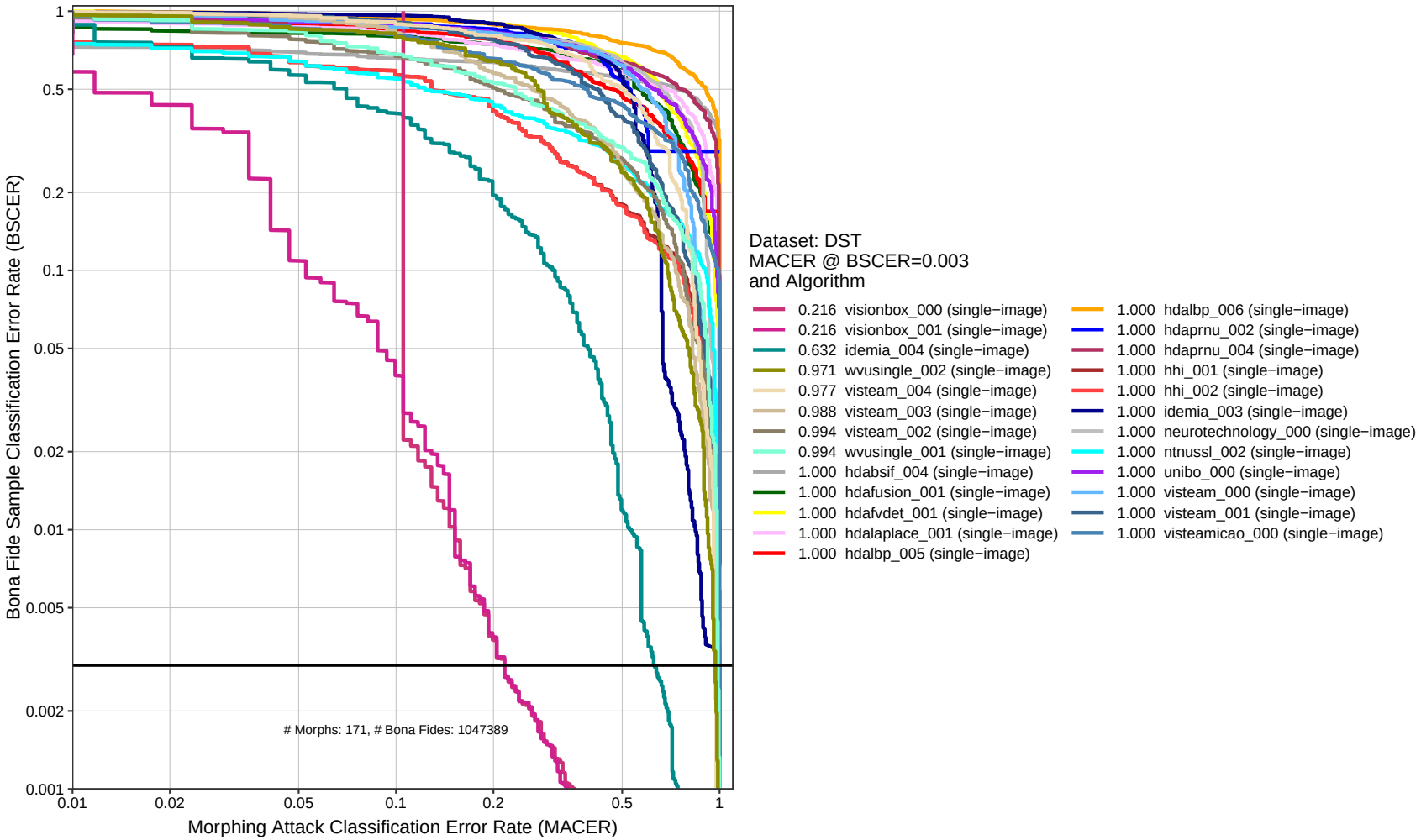


Figure 16: DET plot. This chart plots BSCER as a function of MACER. The x-axis is the rate morphs are not detected and the y-axis is the rate that bona fide images are falsely classified as morphs. The horizontal black line represents BSCER=0.003.

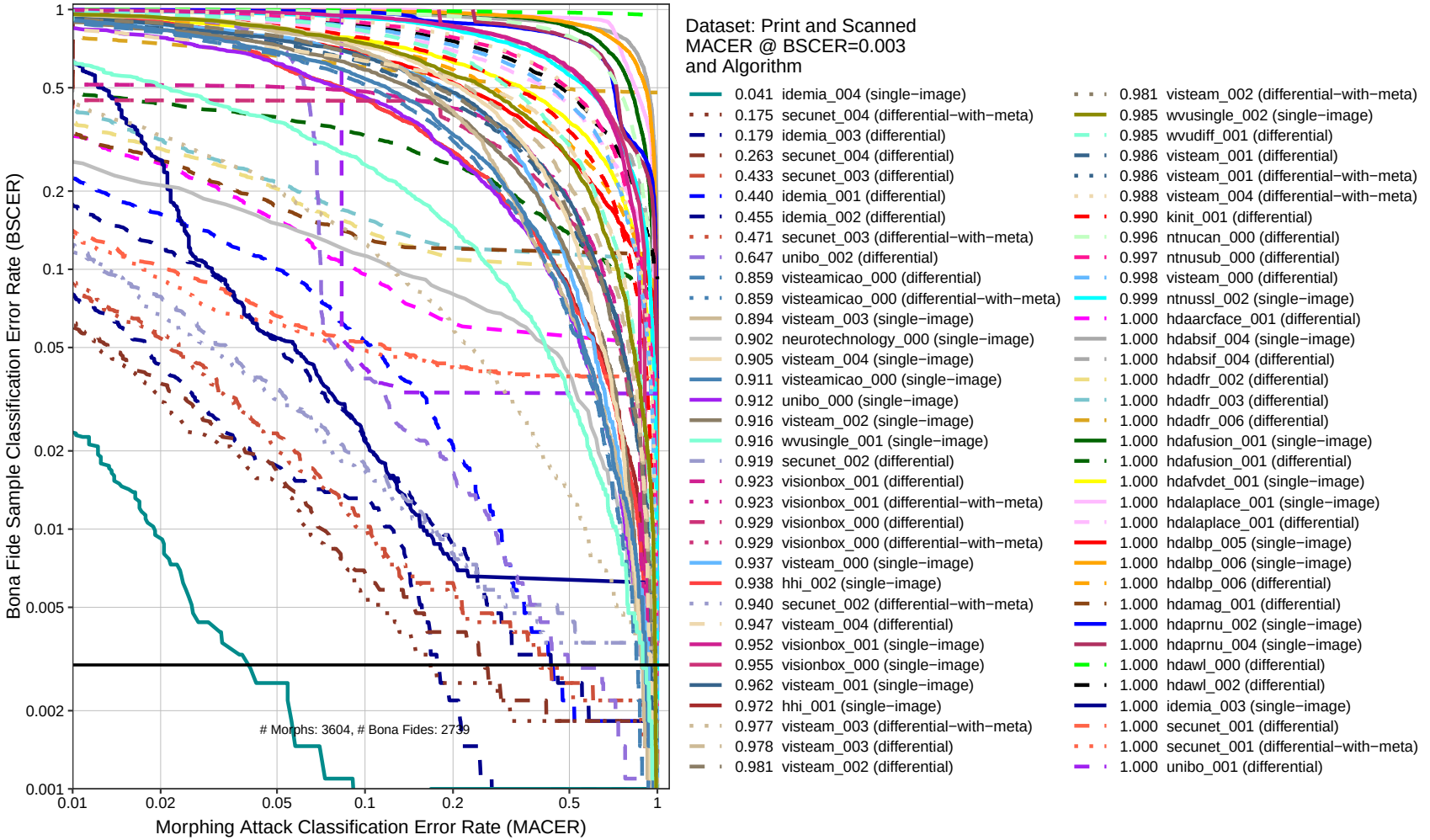


Figure 17: DET plot. This chart plots BSCER as a function of MACER. The x-axis is the rate morphs are not detected and the y-axis is the rate that bona fide images are falsely classified as morphs. The horizontal black line represents BSCER=0.003.

4.7 Impact of Image Resolution

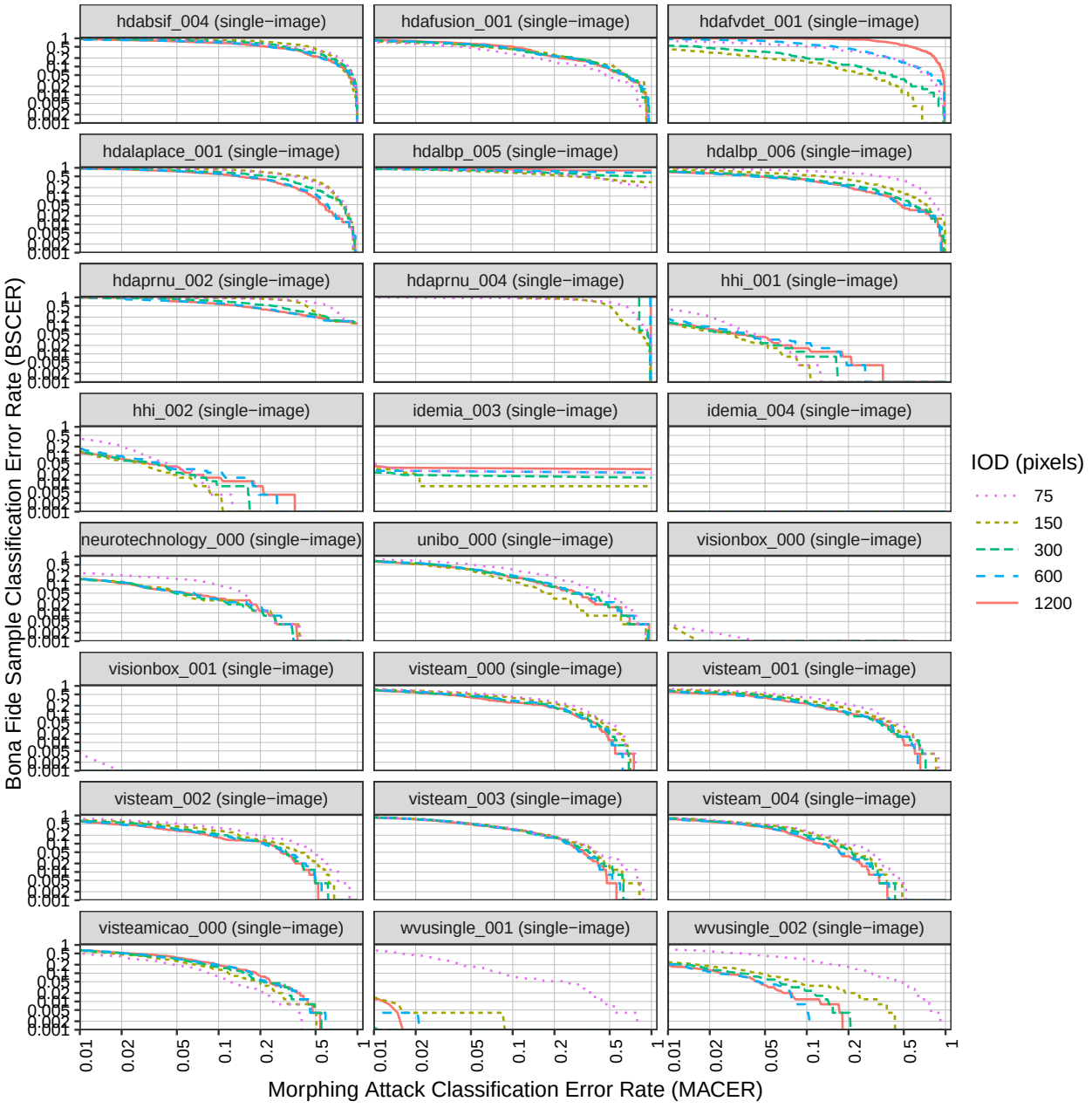


Figure 18: The DET curves show single-image morph detection error rates for different image resolutions, reported as interocular distance (IOD) or the distance between the eyes. Note that these DET curves do not show MACER and BSCER at fixed morph detection score thresholds between different image resolutions. Please refer to Figures 19 and 20 for assessments of MACER and BSCER as a function of score threshold, respectively. For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

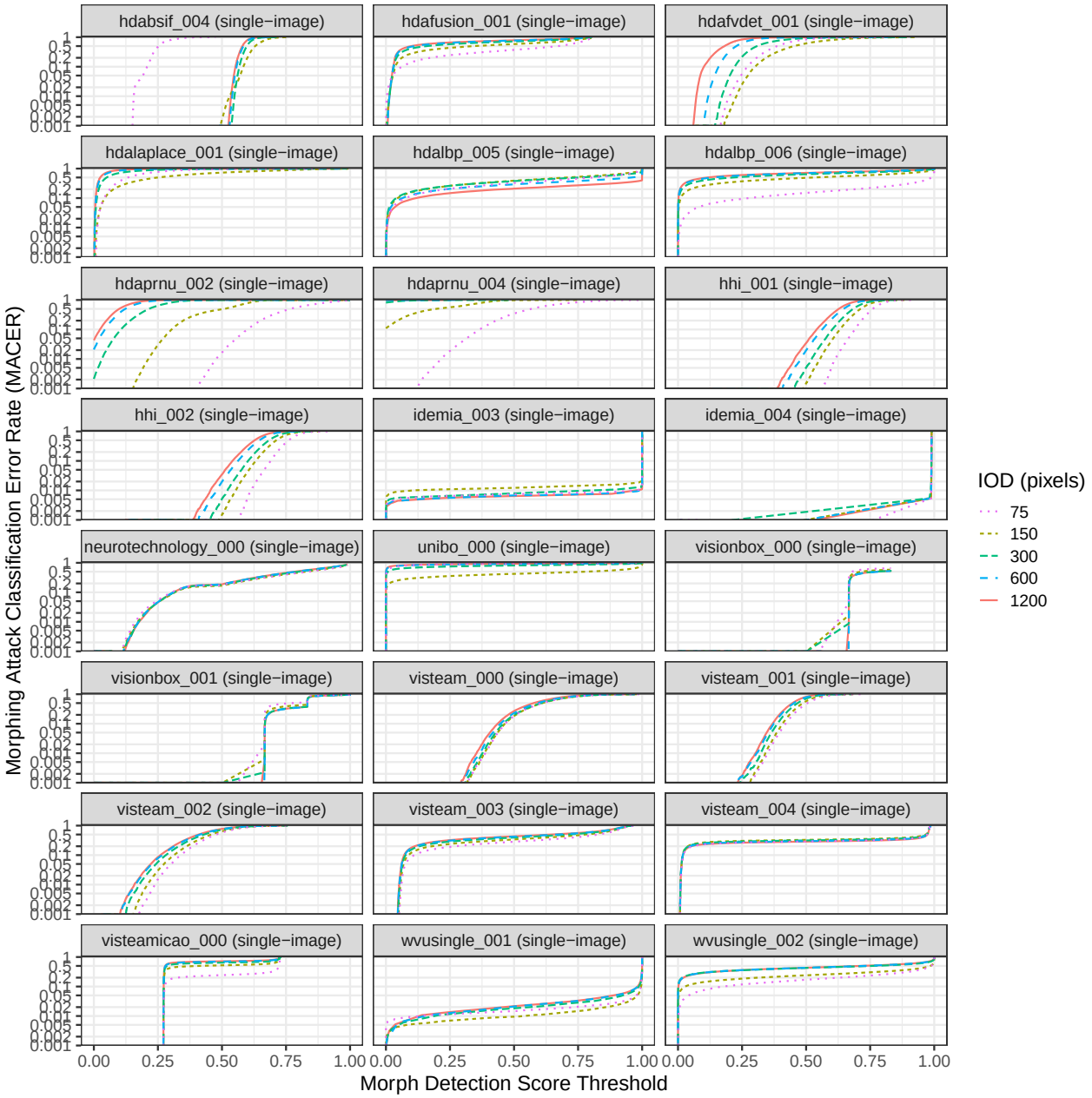


Figure 19: The curves show MACER (or morph miss rate) as a function of morph detection score threshold for different image resolutions, reported as interocular distance (IOD), the distance between the eyes. For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

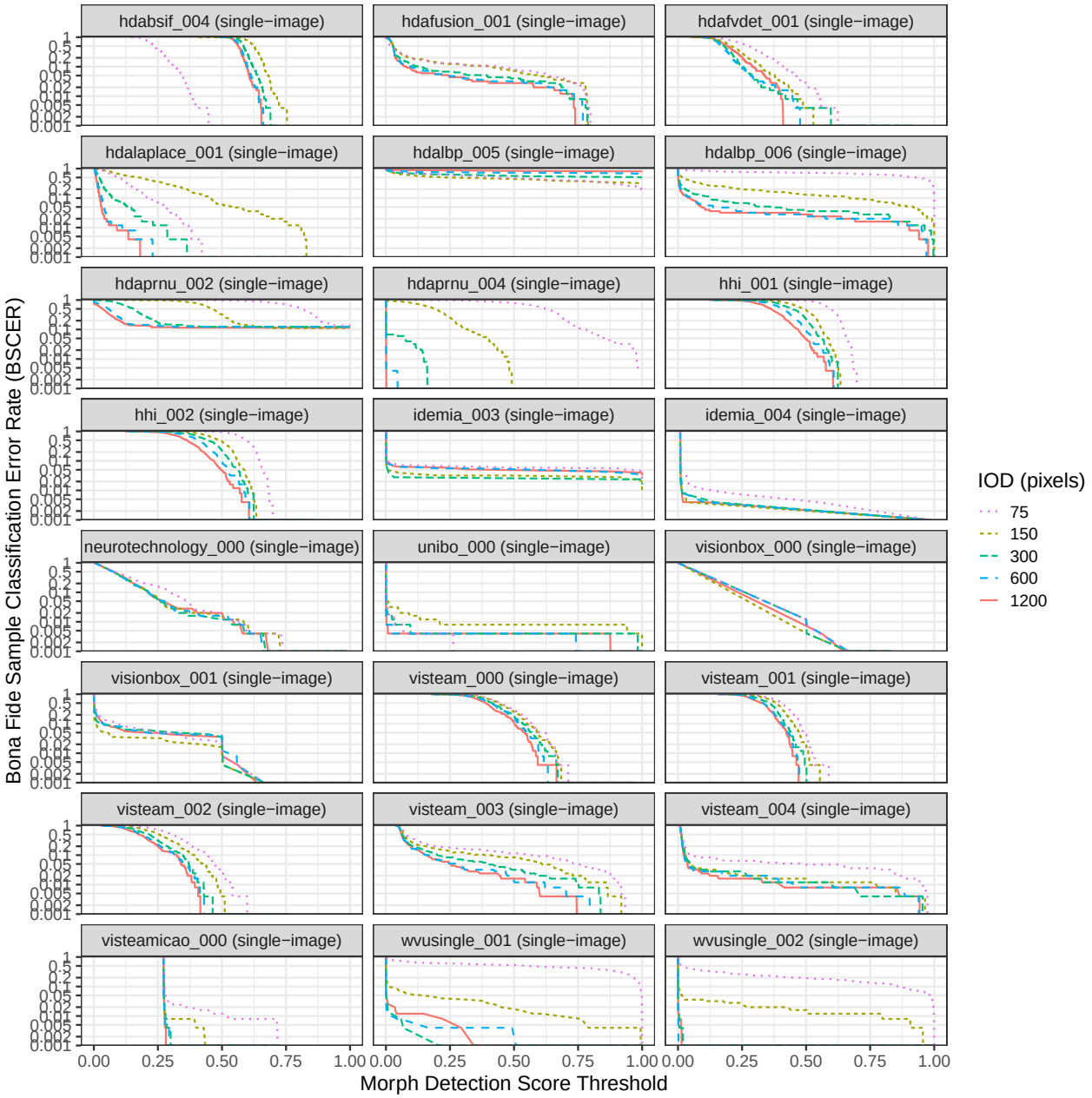


Figure 20: The curves show BSCER (or false detection rate) as a function of morph detection score threshold for different image resolutions, reported as interocular distance (IOD), the distance between the eyes. For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

4.8 BSCER and MACER Calibration

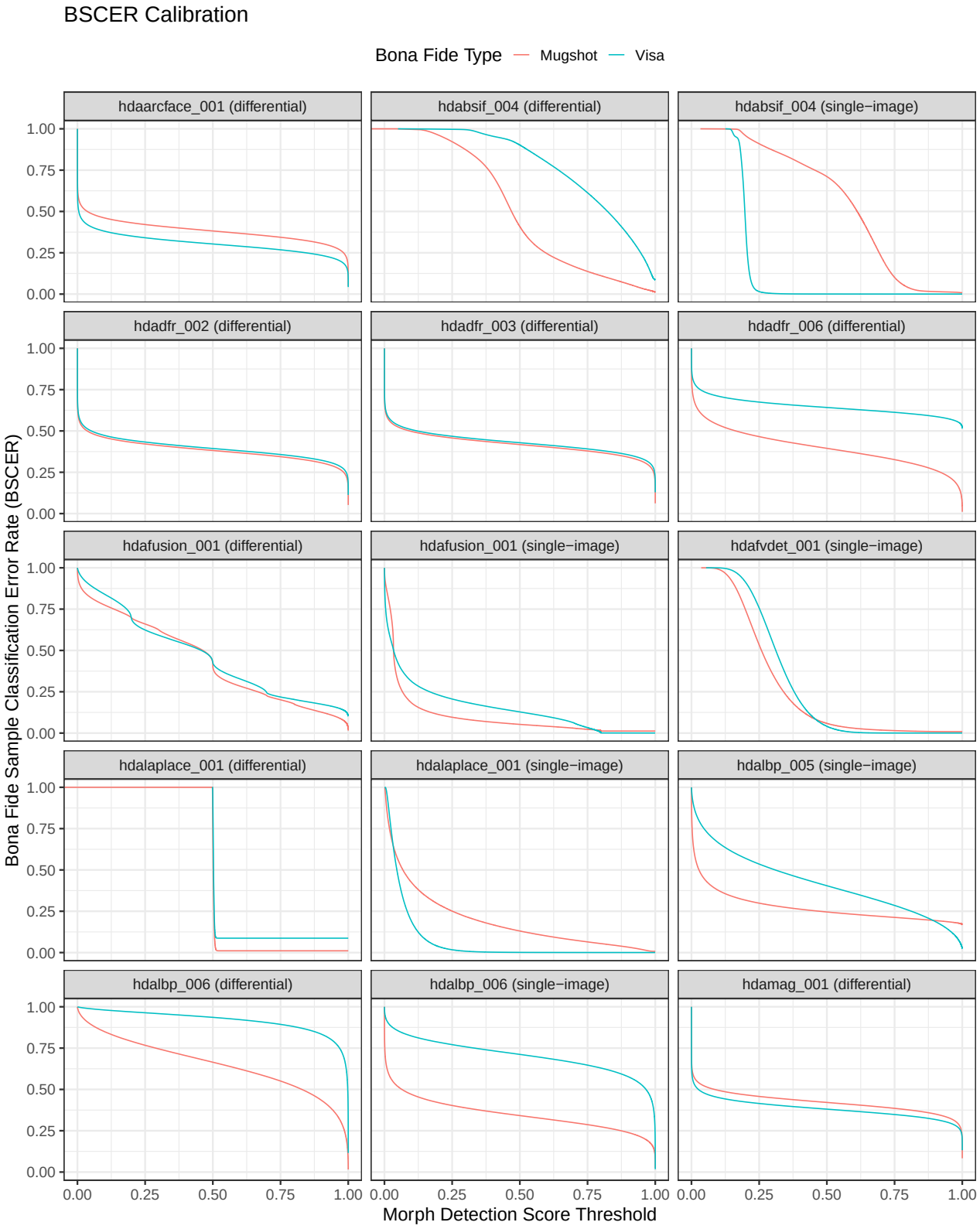


Figure 21: The BSCER calibration curves show BSCER (or false detection rate) vs. morph detection score threshold. Separate curves appear for mugshot and visa images. For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

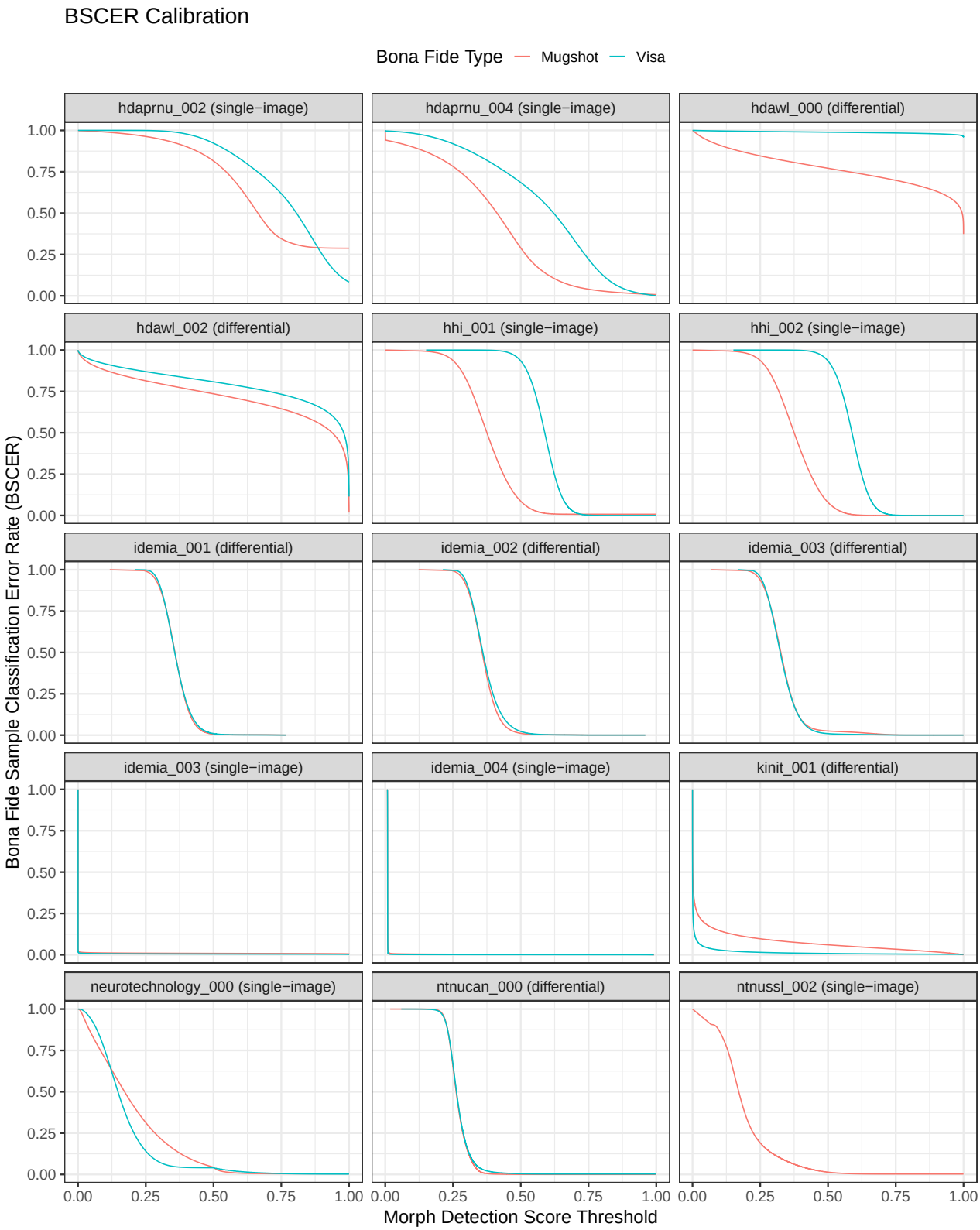


Figure 22: The BSCER calibration curves show BSCER (or false detection rate) vs. morph detection score threshold. Separate curves appear for mugshot and visa images. For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

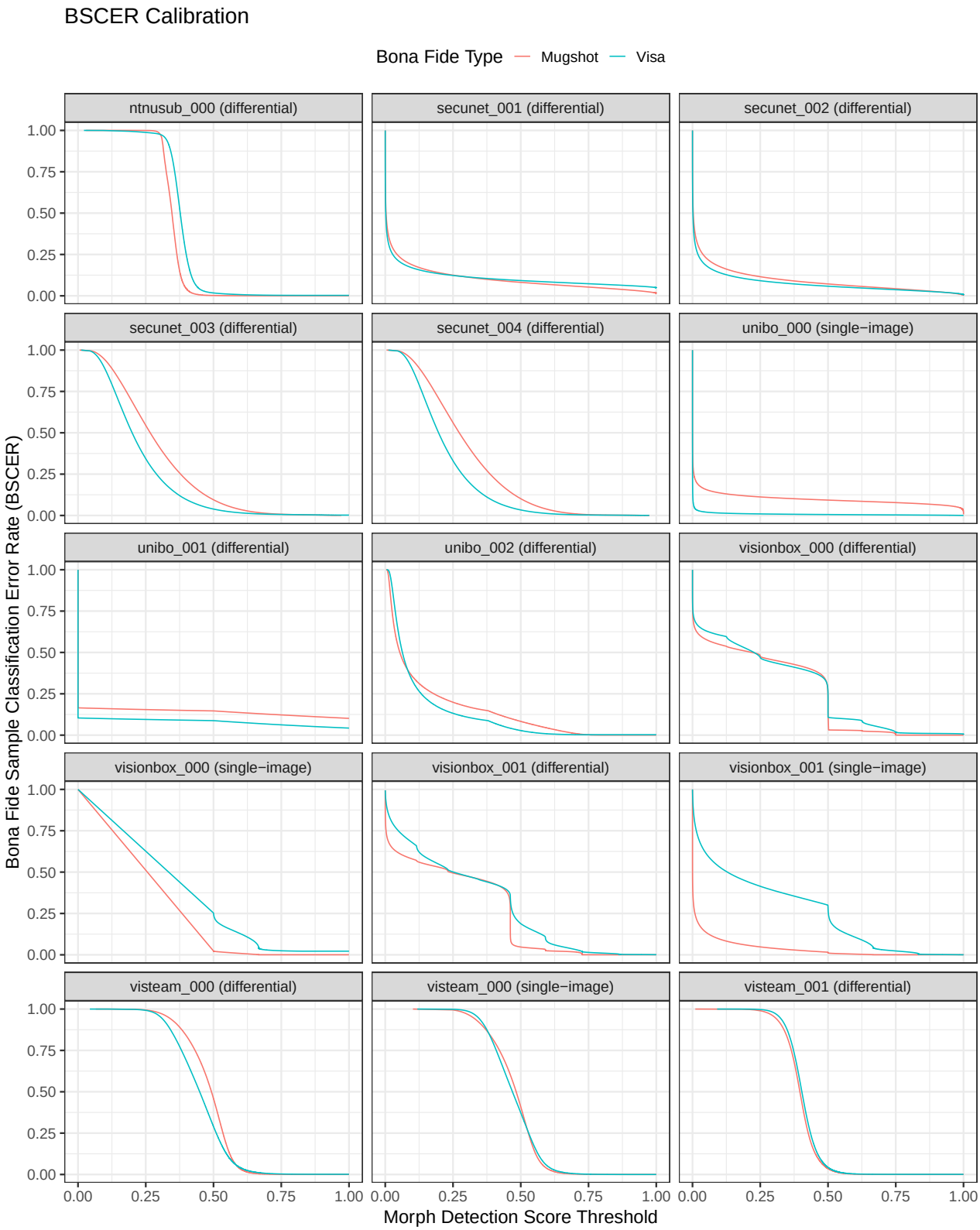


Figure 23: The BSCER calibration curves show BSCER (or false detection rate) vs. morph detection score threshold. Separate curves appear for mugshot and visa images. For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

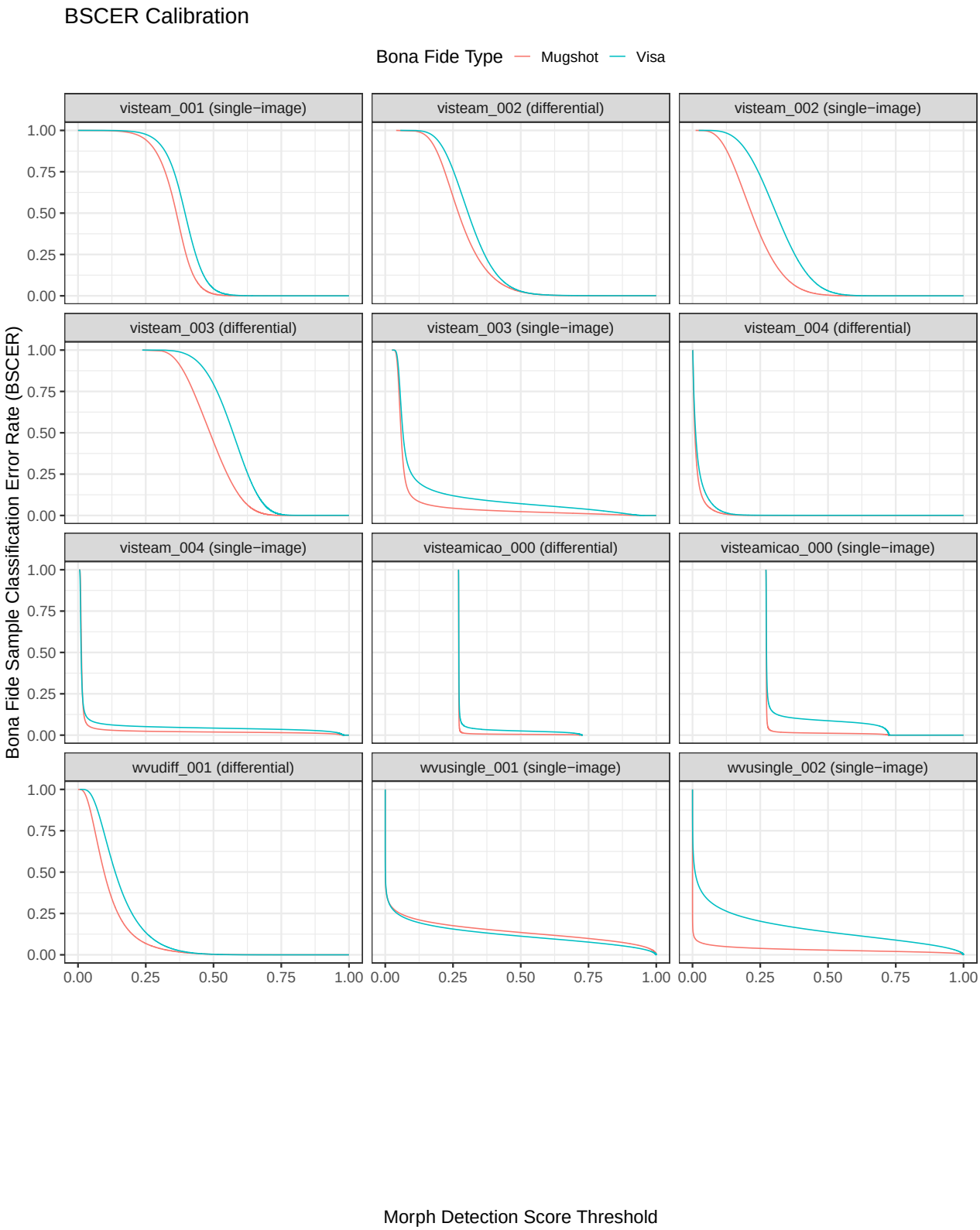


Figure 24: The BSCER calibration curves show BSCER (or false detection rate) vs. morph detection score threshold. Separate curves appear for mugshot and visa images. For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

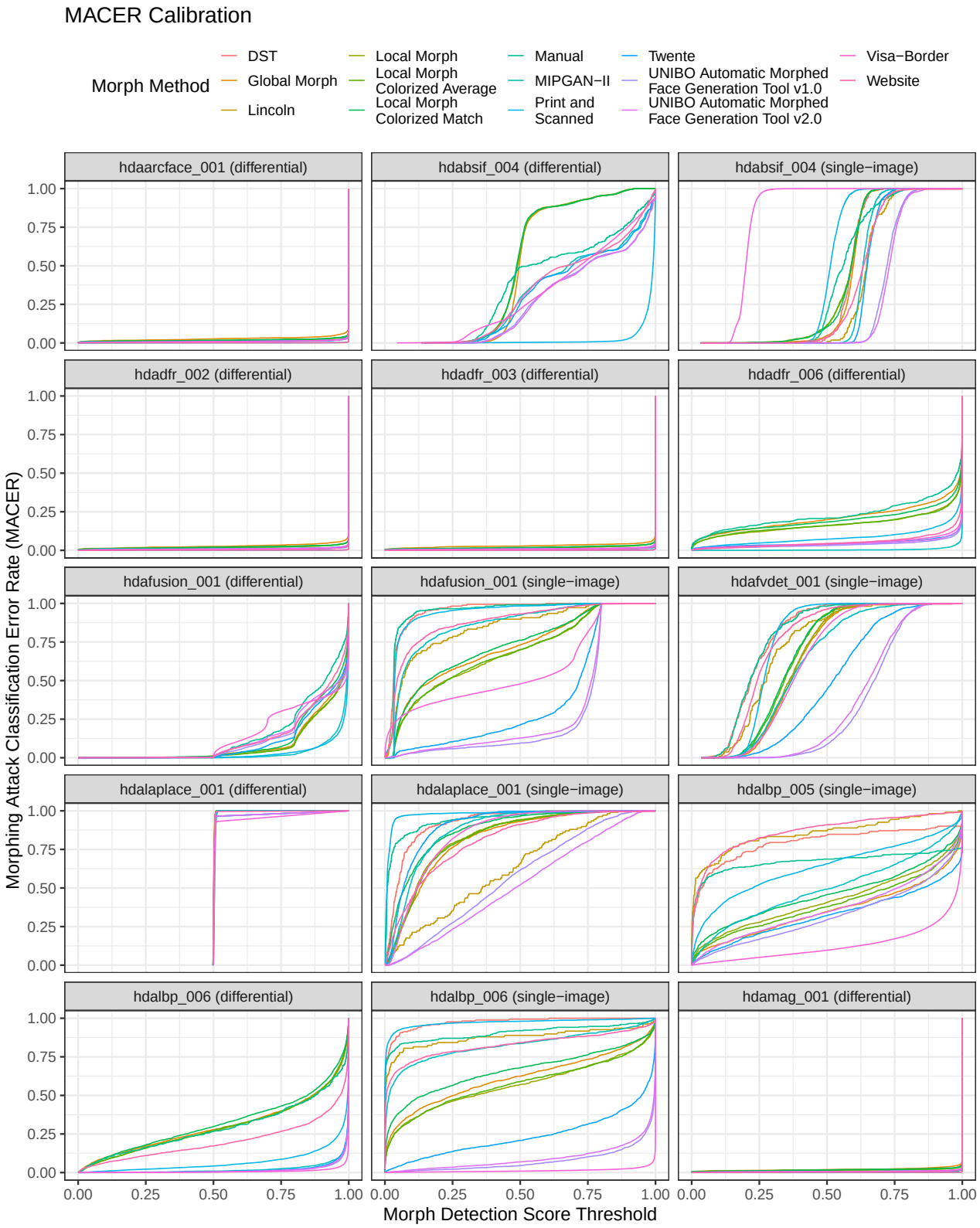


Figure 25: The MACER calibration curves show MACER (or morph miss rate) vs. morph detection score threshold. Separate curves appear for each morph dataset. For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

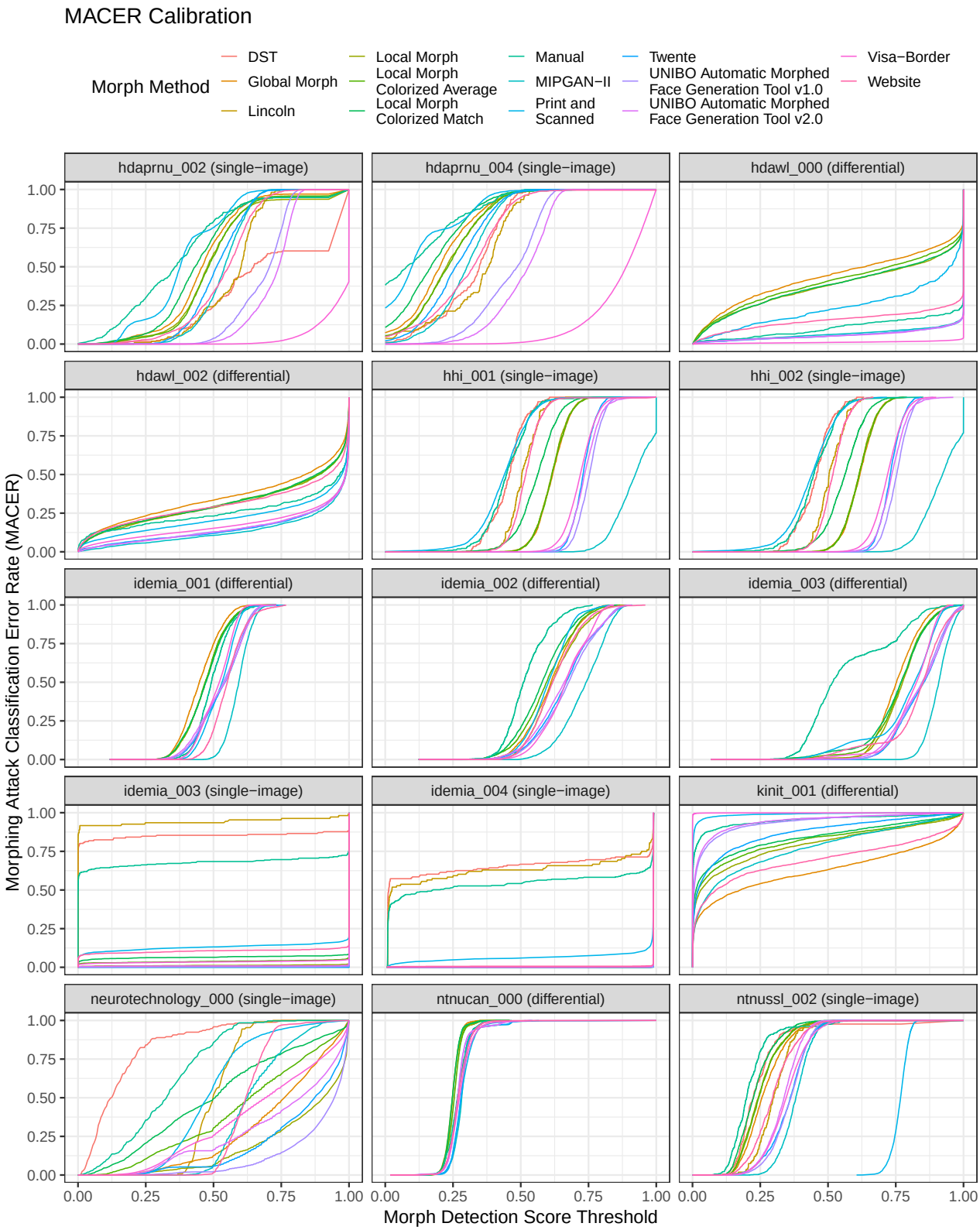


Figure 26: The MACER calibration curves show MACER (or morph miss rate) vs. morph detection score threshold. Separate curves appear for each morph dataset. For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

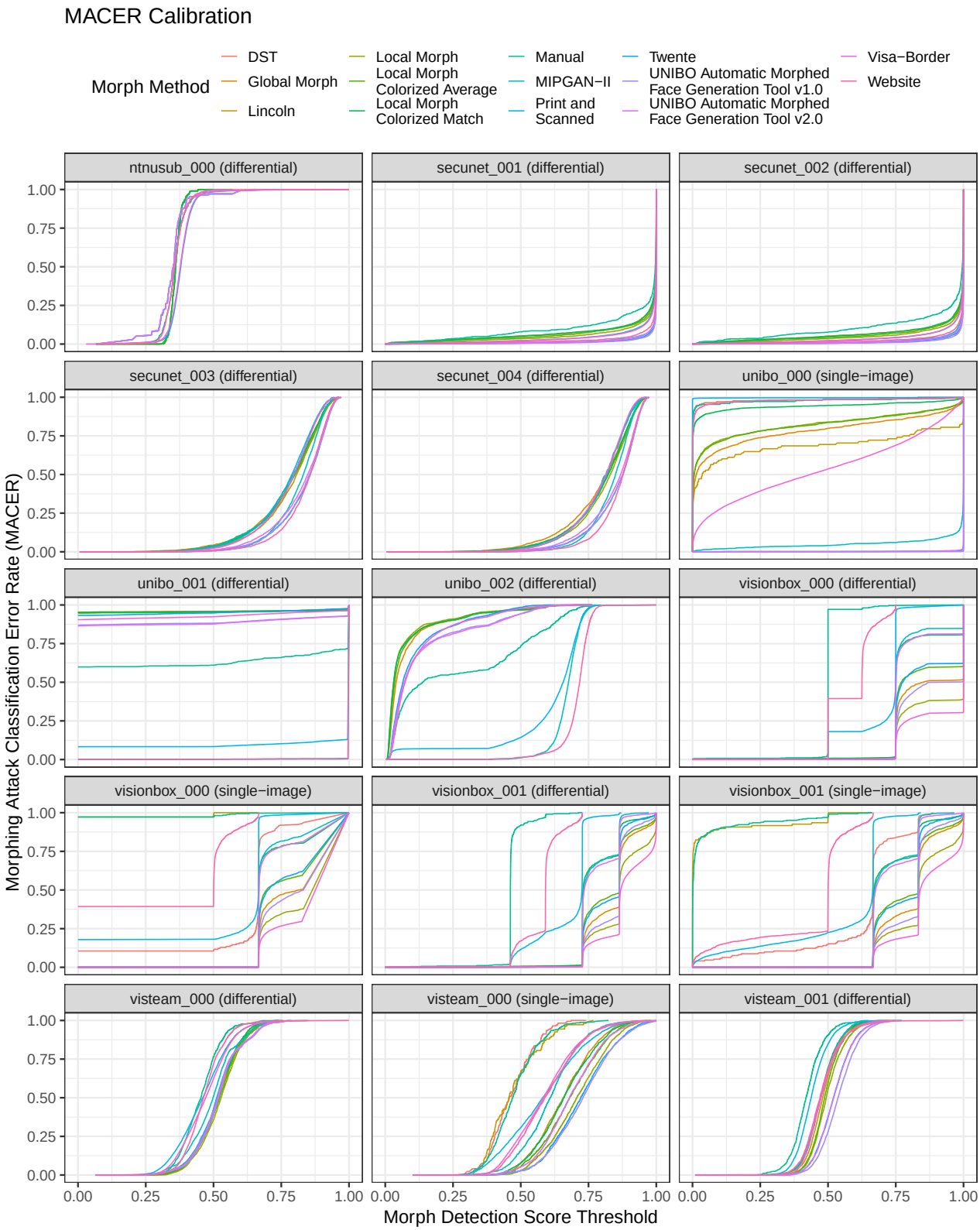


Figure 27: The MACER calibration curves show MACER (or morph miss rate) vs. morph detection score threshold. Separate curves appear for each morph dataset. For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

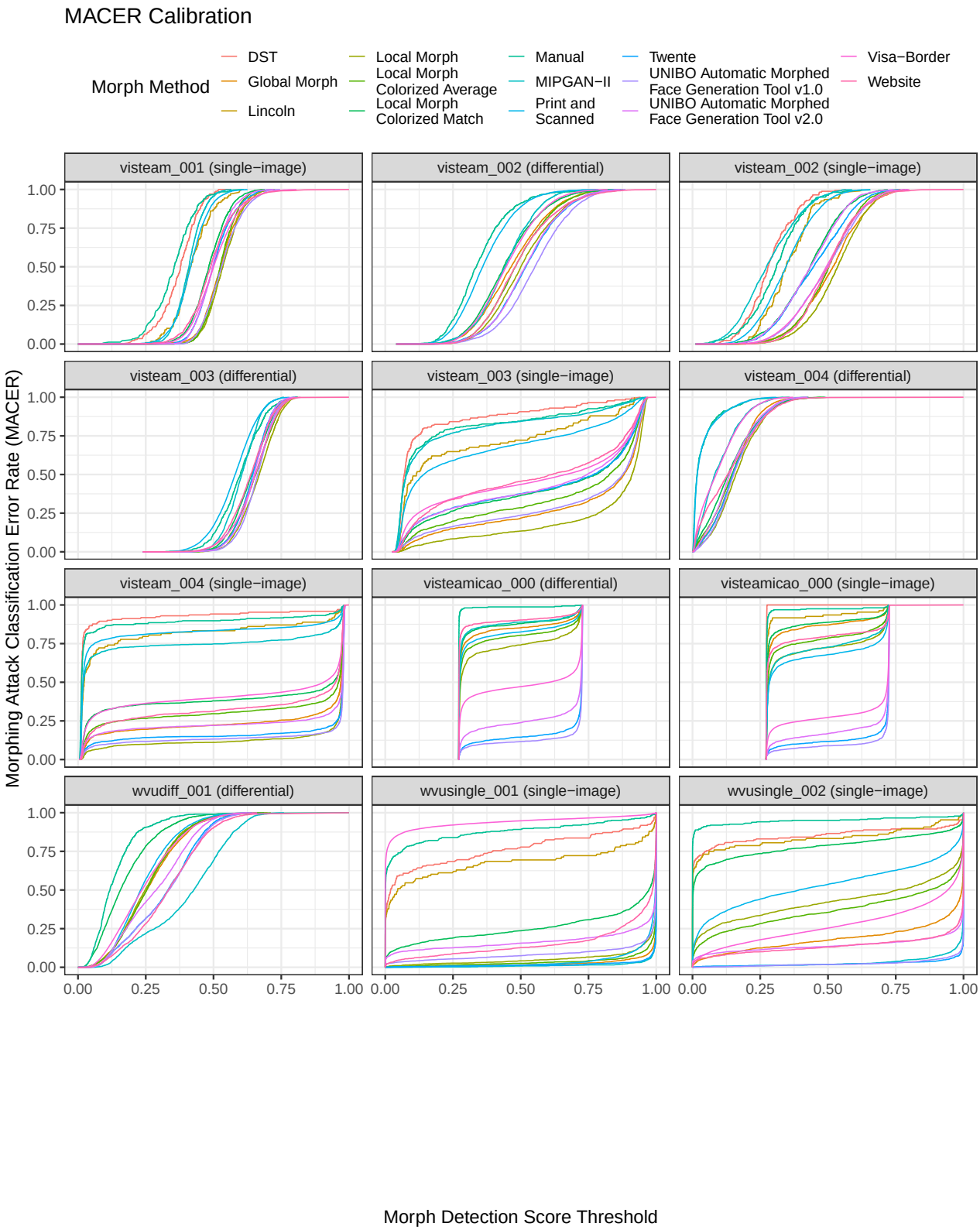


Figure 28: The MACER calibration curves show MACER (or morph miss rate) vs. morph detection score threshold. Separate curves appear for each morph dataset. For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

4.9 Morph Detection Scores vs. Elapsed Time (Two-image differential)

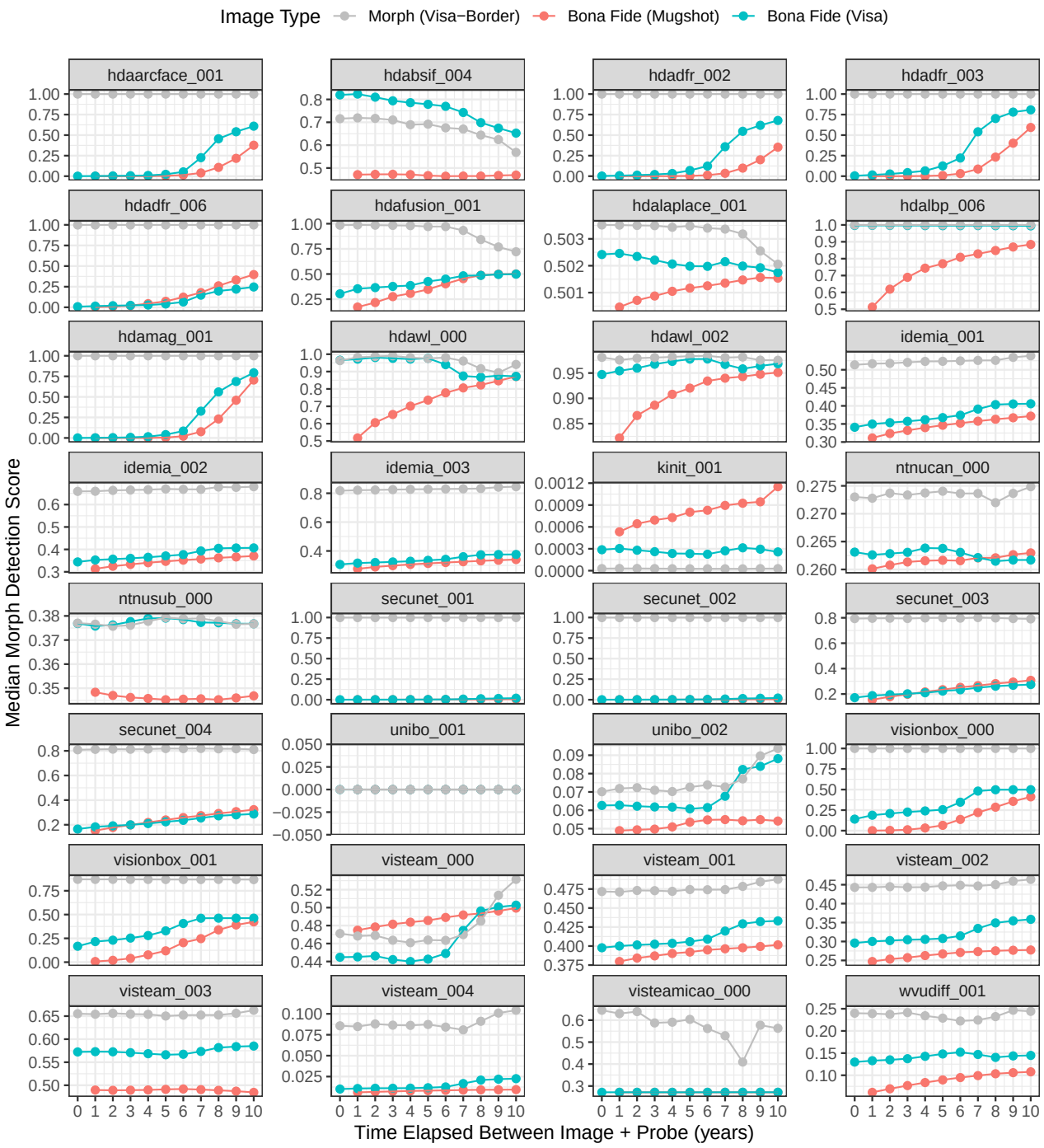


Figure 29: For the visa and mugshot datasets + probes used to evaluate differential MAD, this figure shows median morph detection score as a function of the time elapsed between the collection of the bona fide image and the live capture probe. For reference, median morph detection scores for the visa-border morph dataset is also shown. Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

4.10 Impact of Subject Alpha

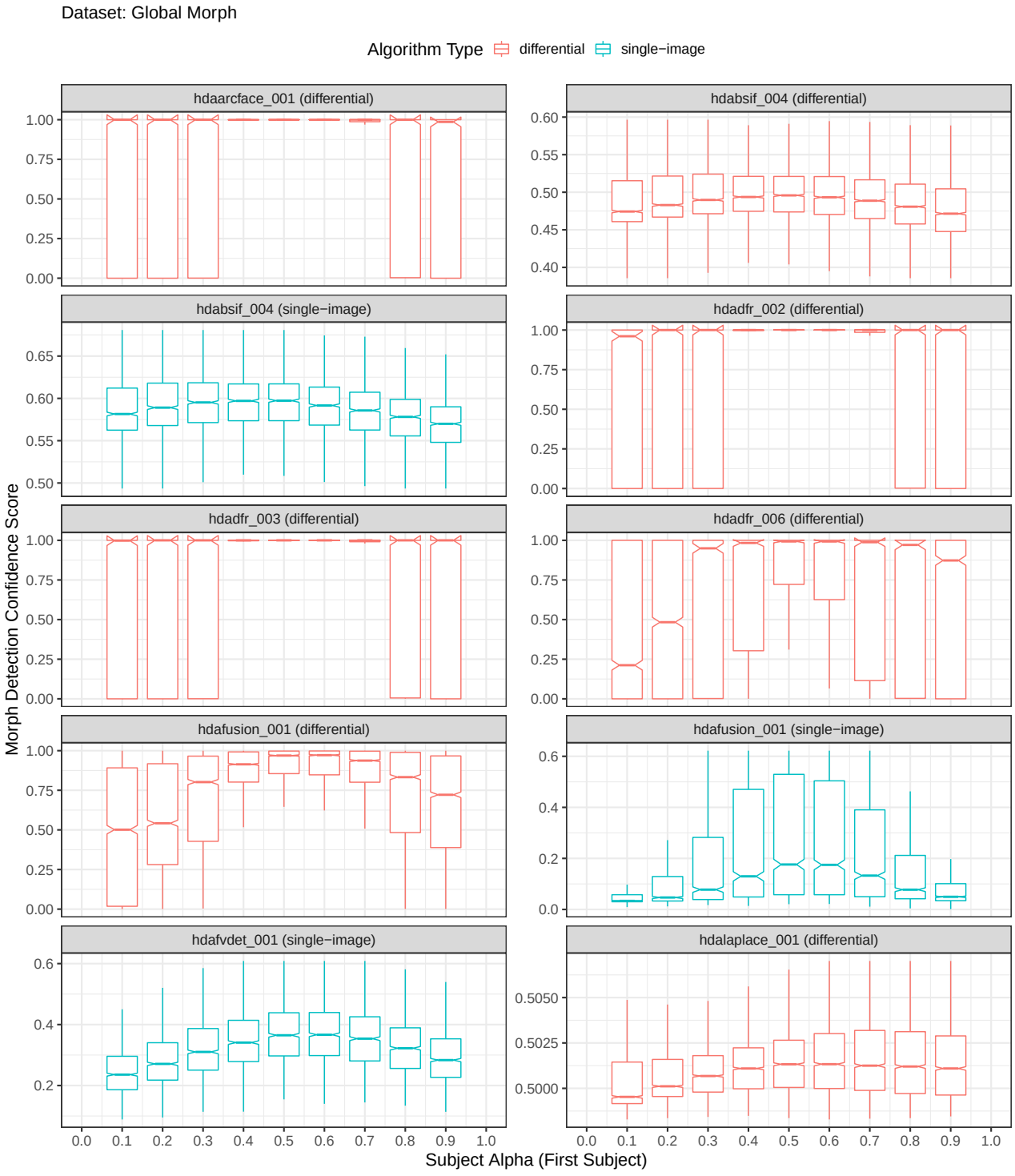


Figure 30: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

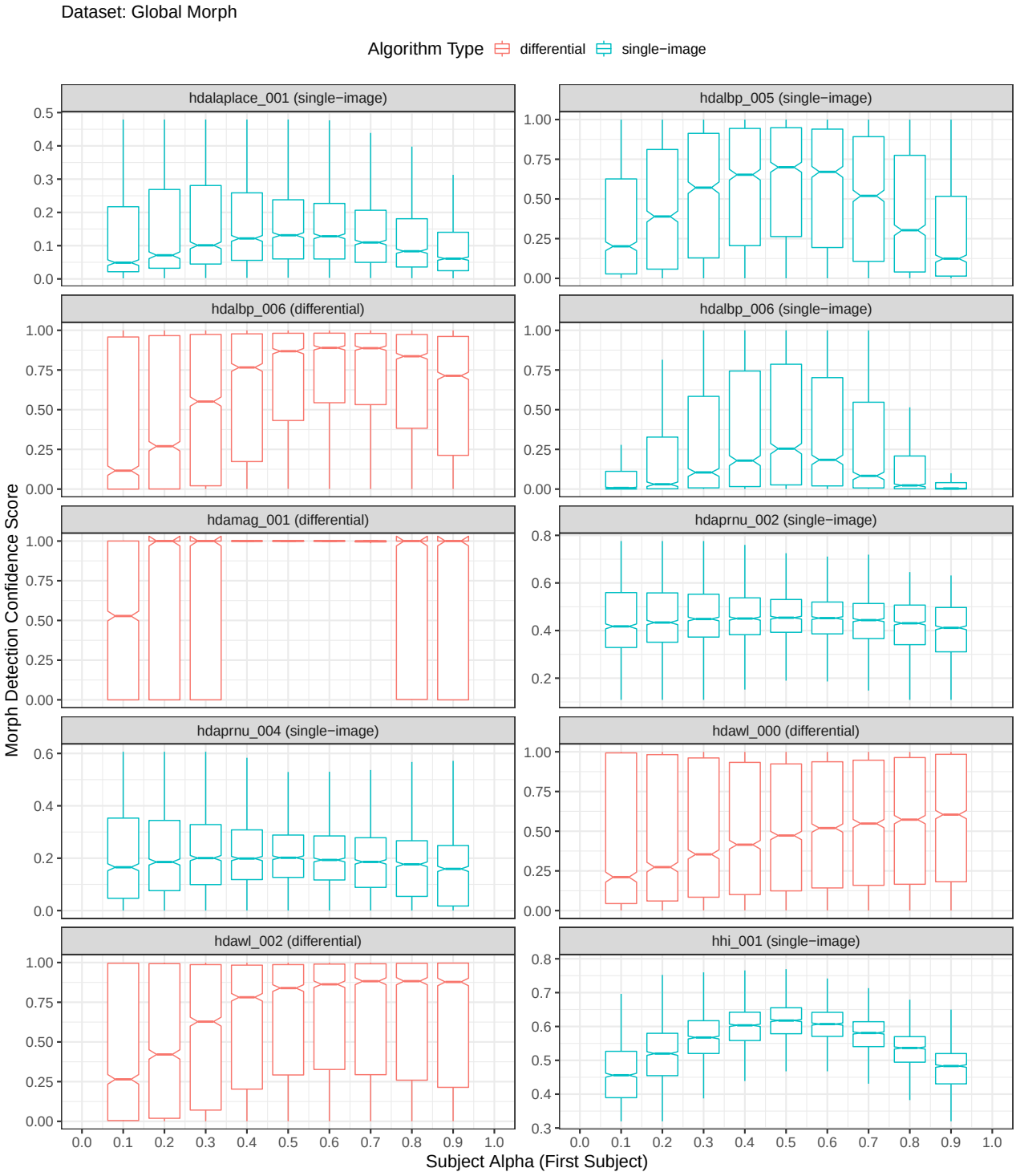


Figure 31: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

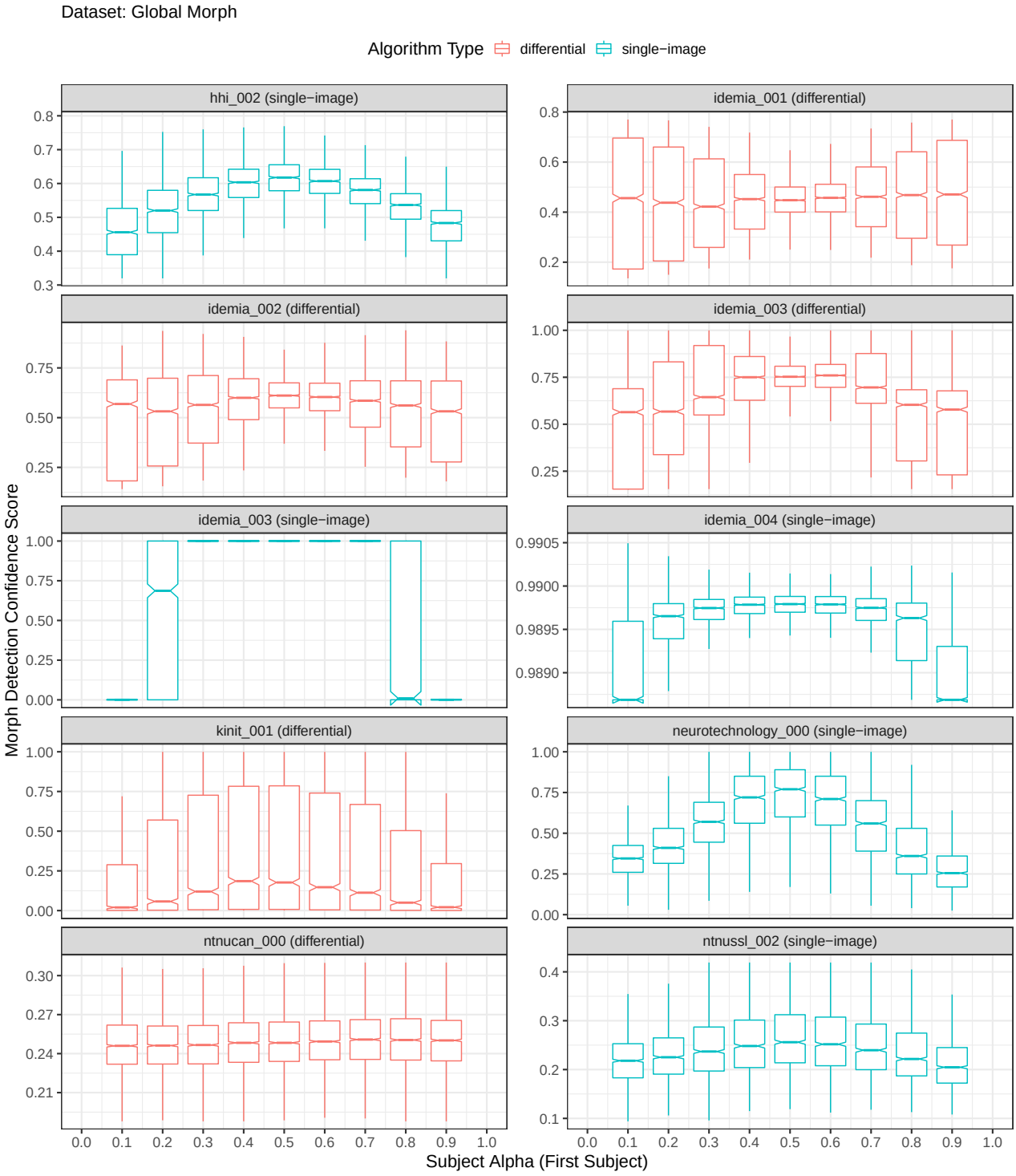


Figure 32: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

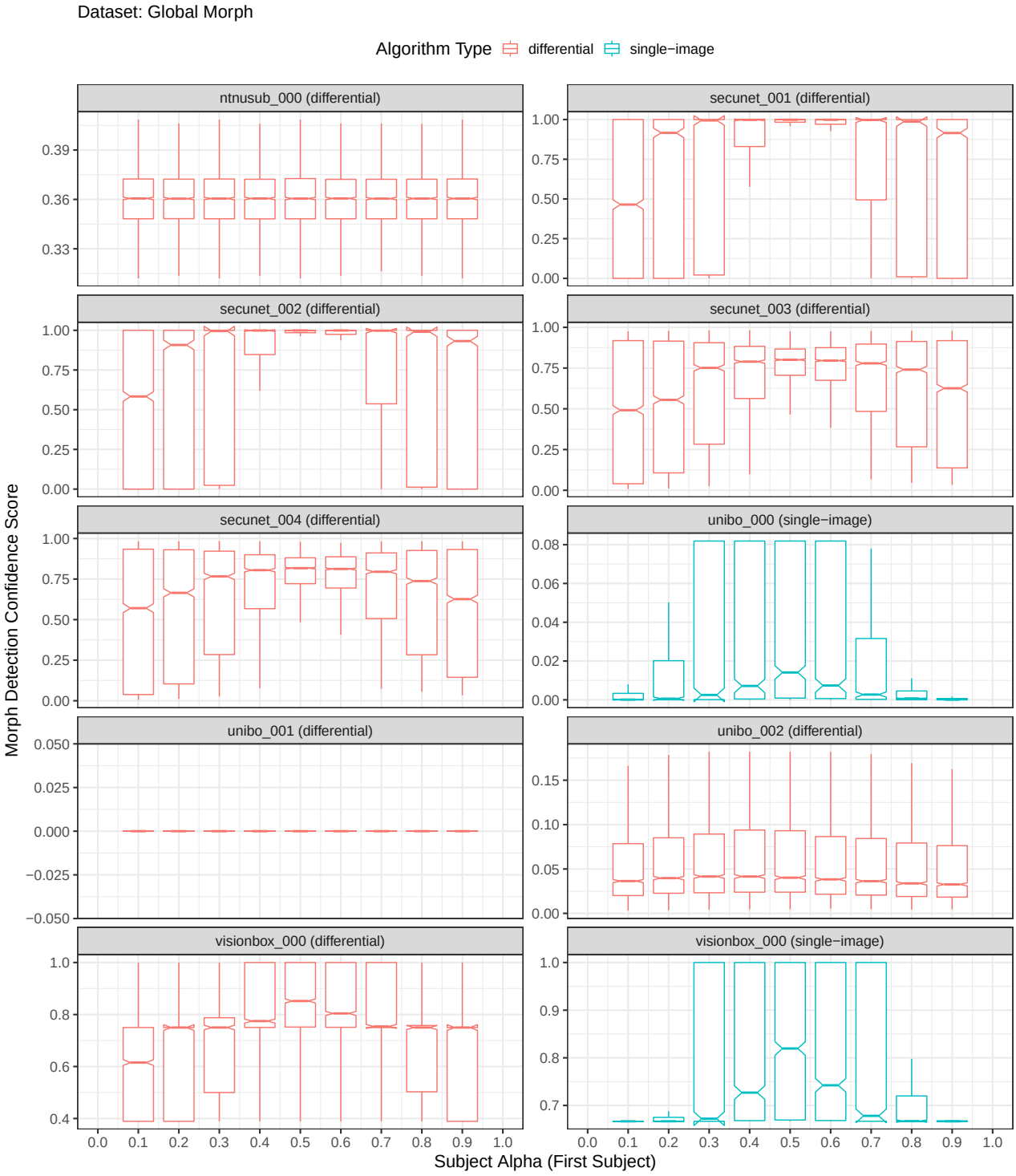


Figure 33: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

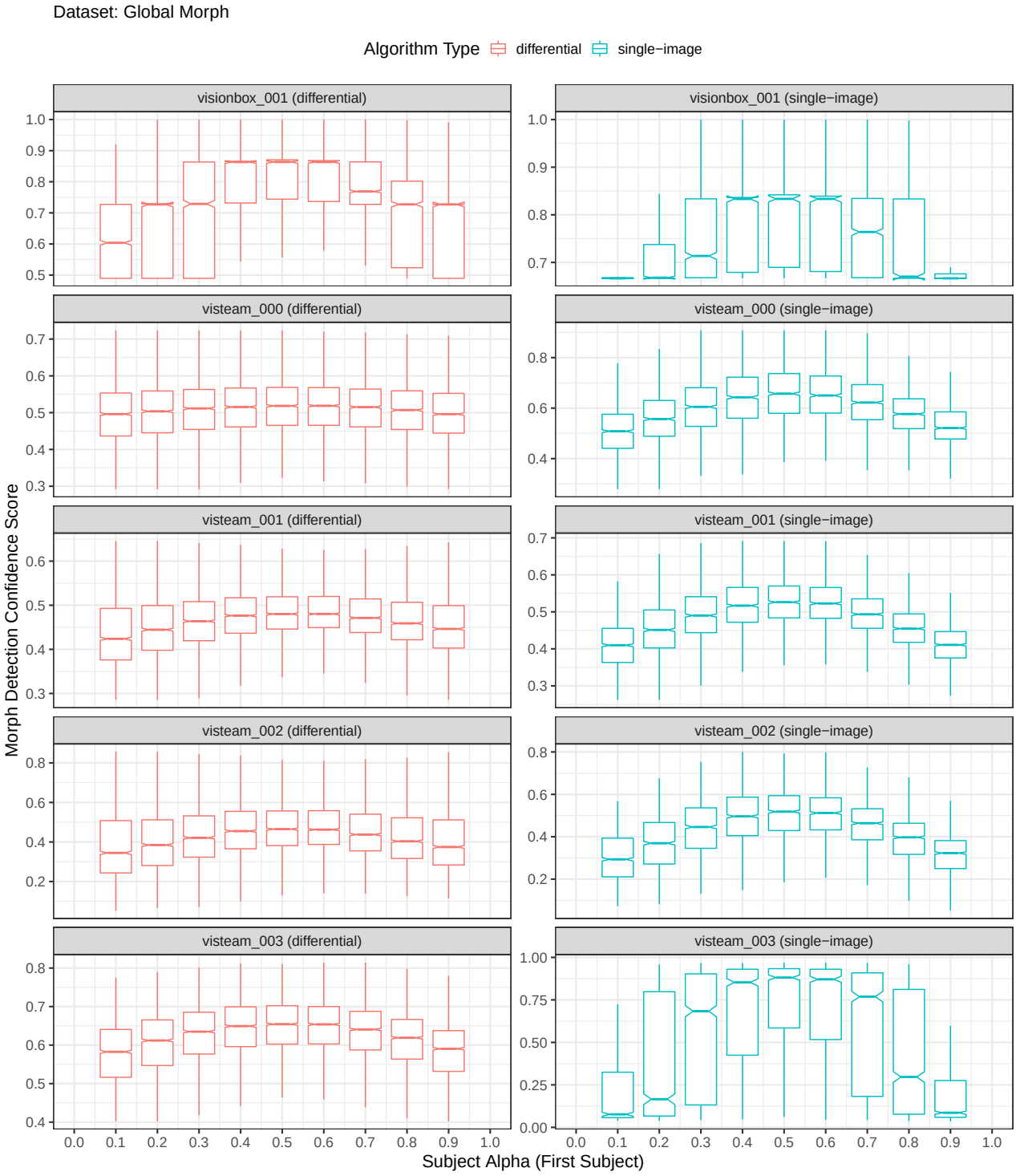


Figure 34: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

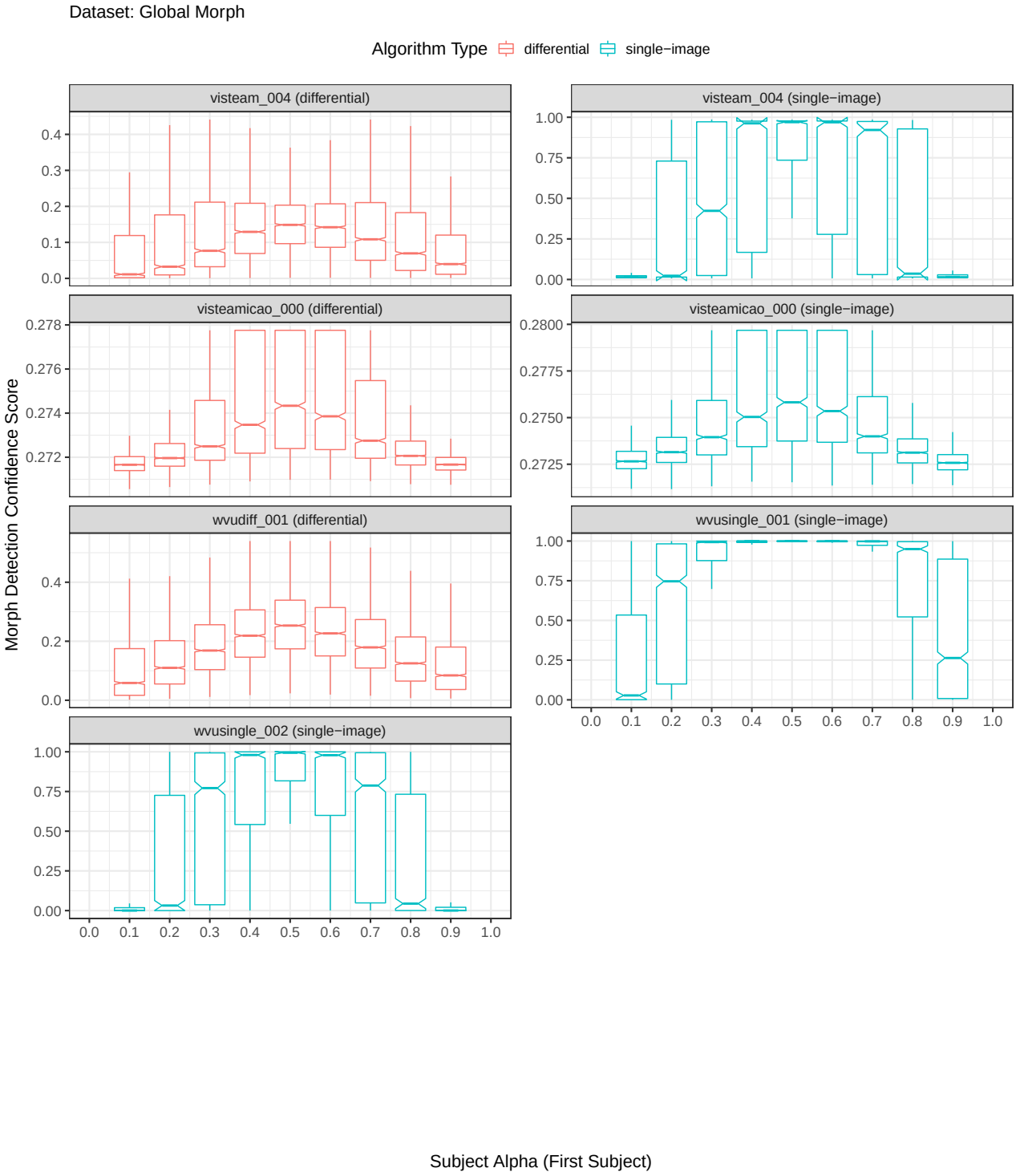


Figure 35: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

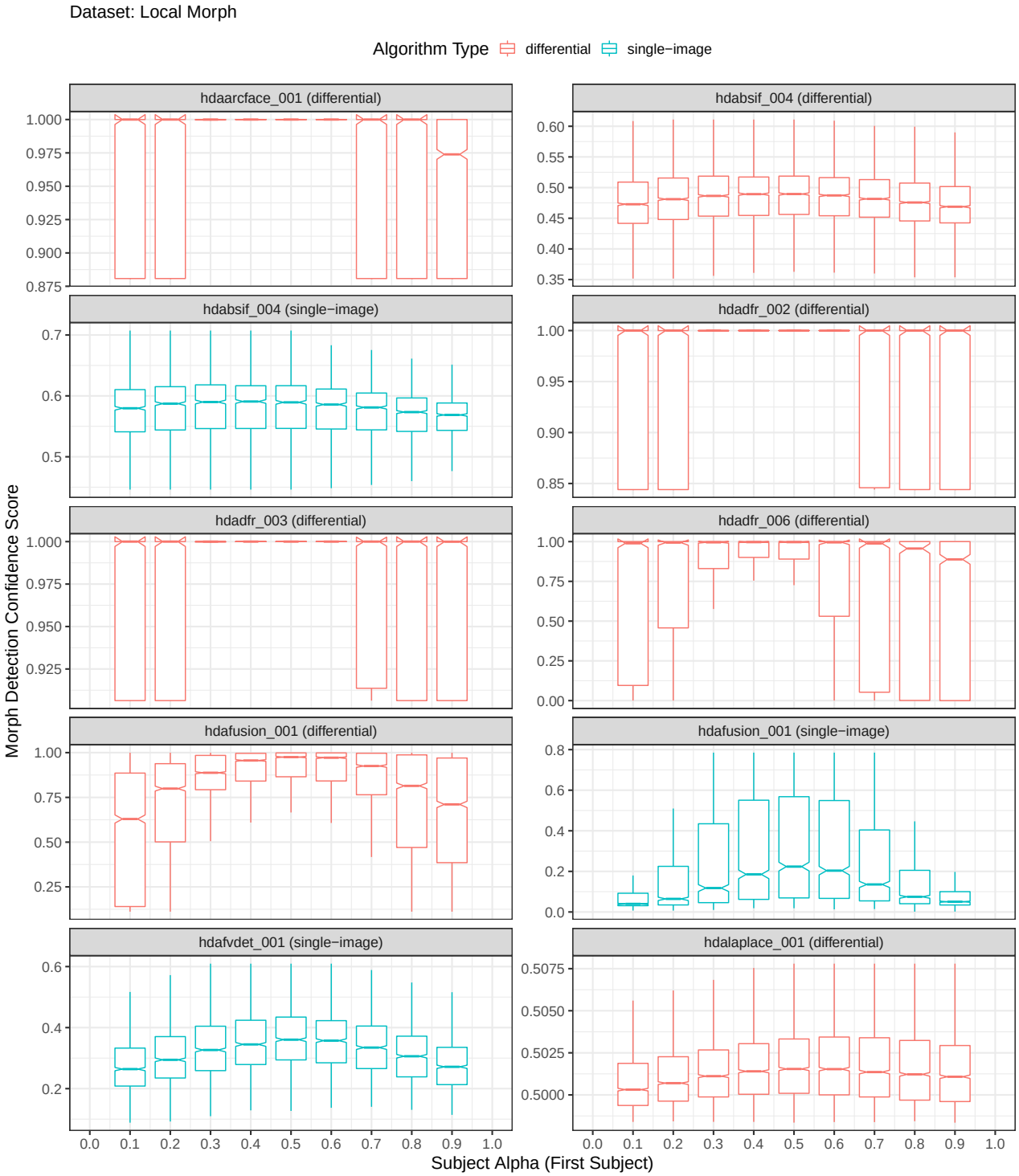


Figure 36: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

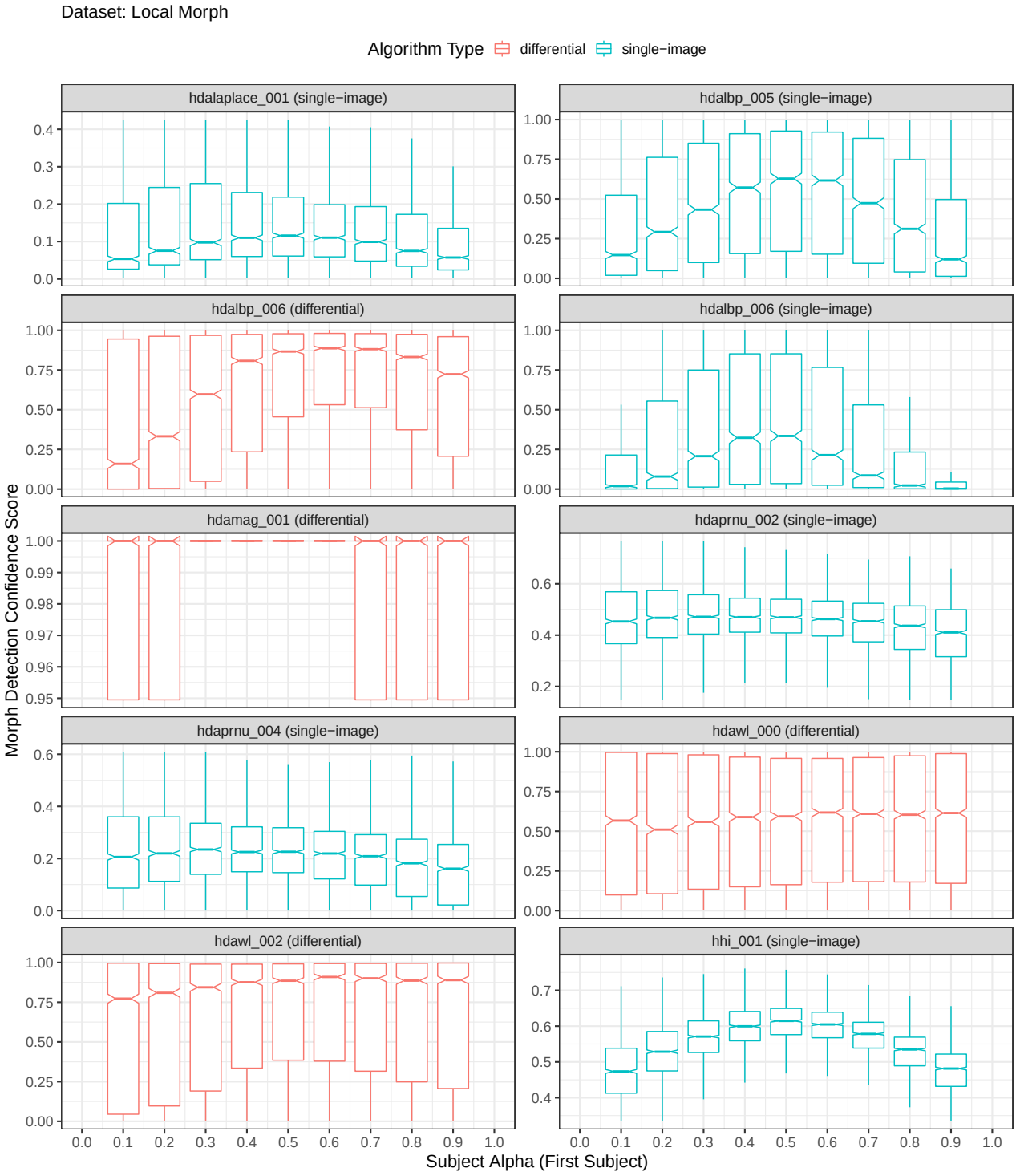


Figure 37: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

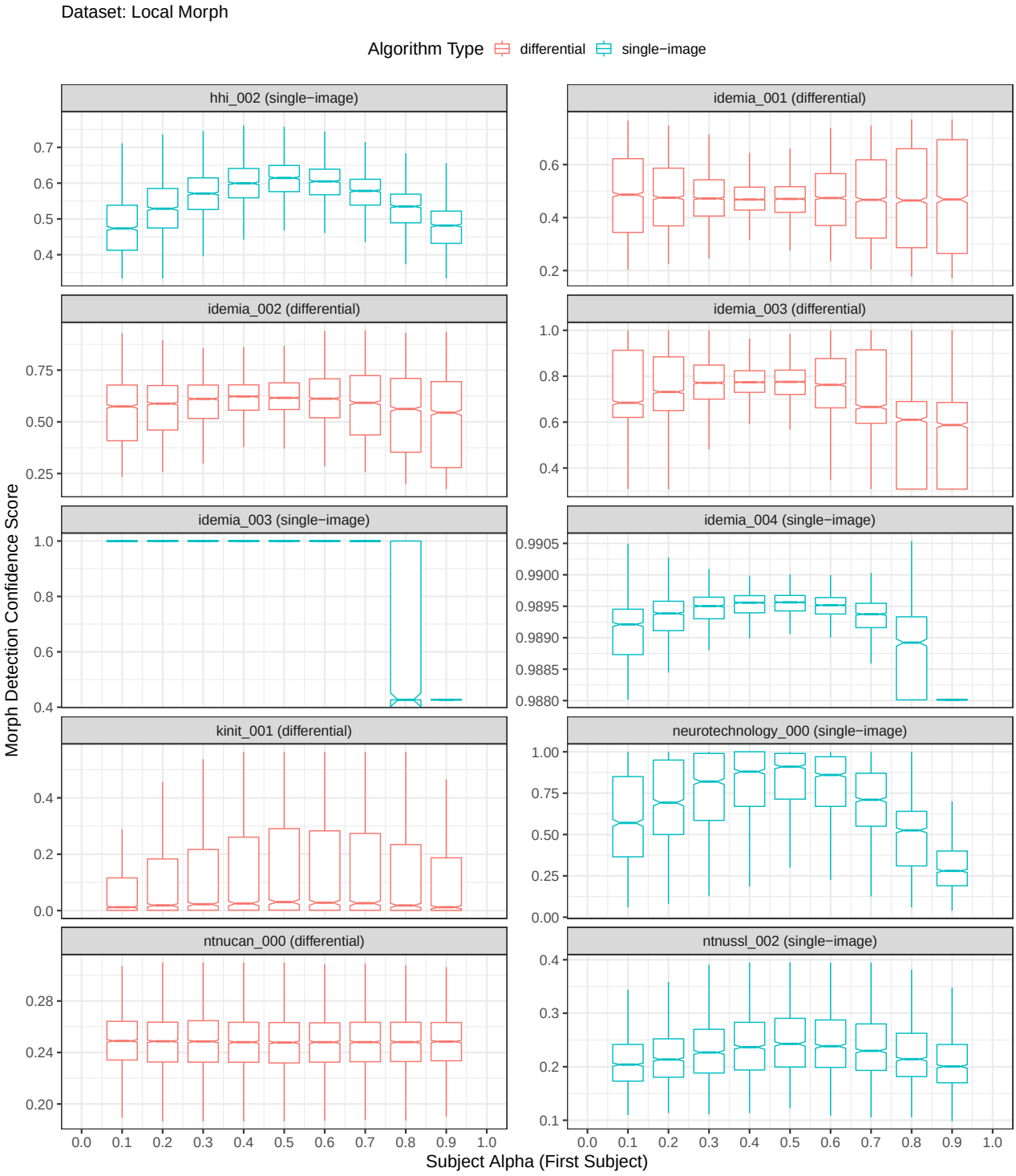


Figure 38: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

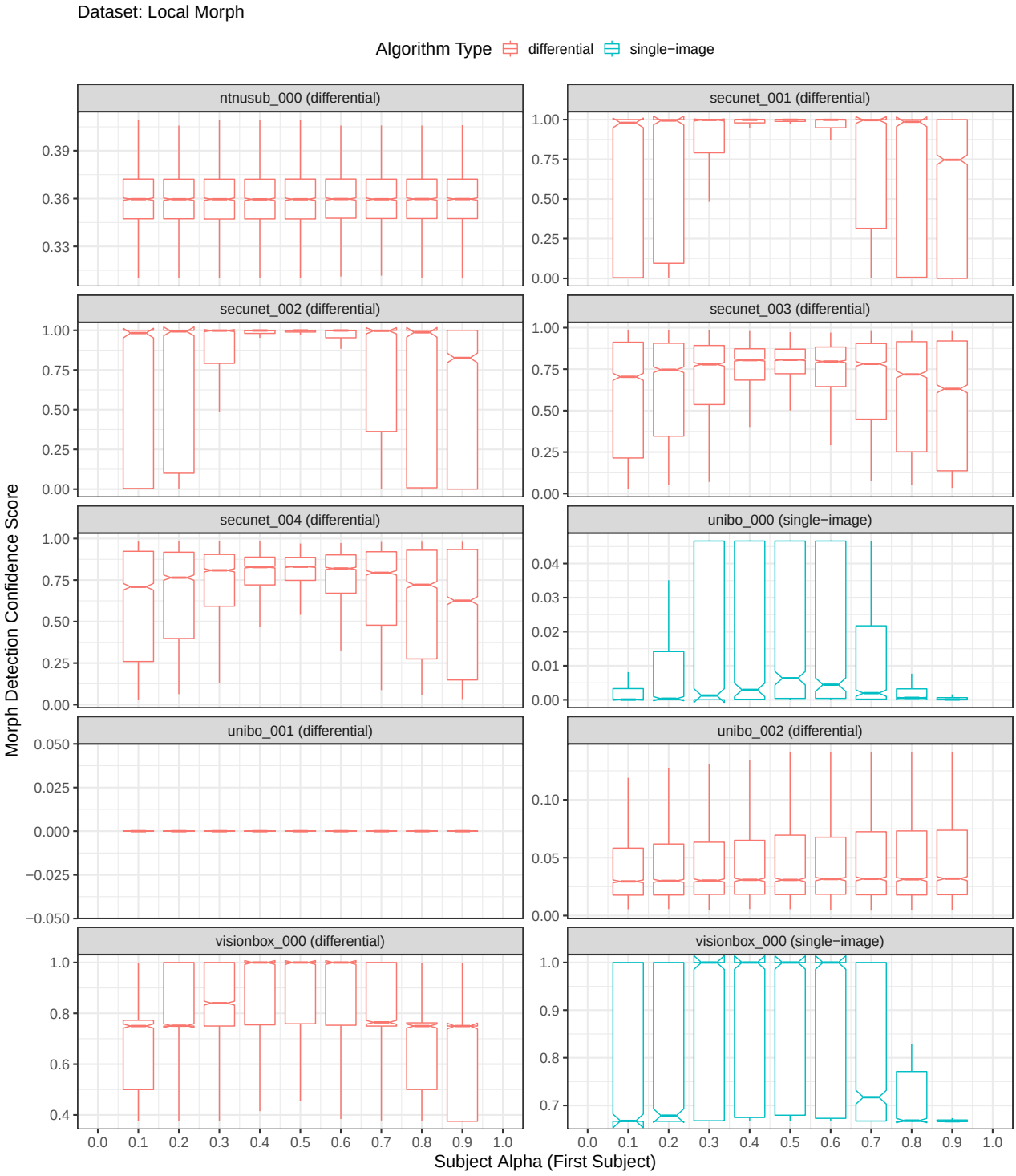


Figure 39: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

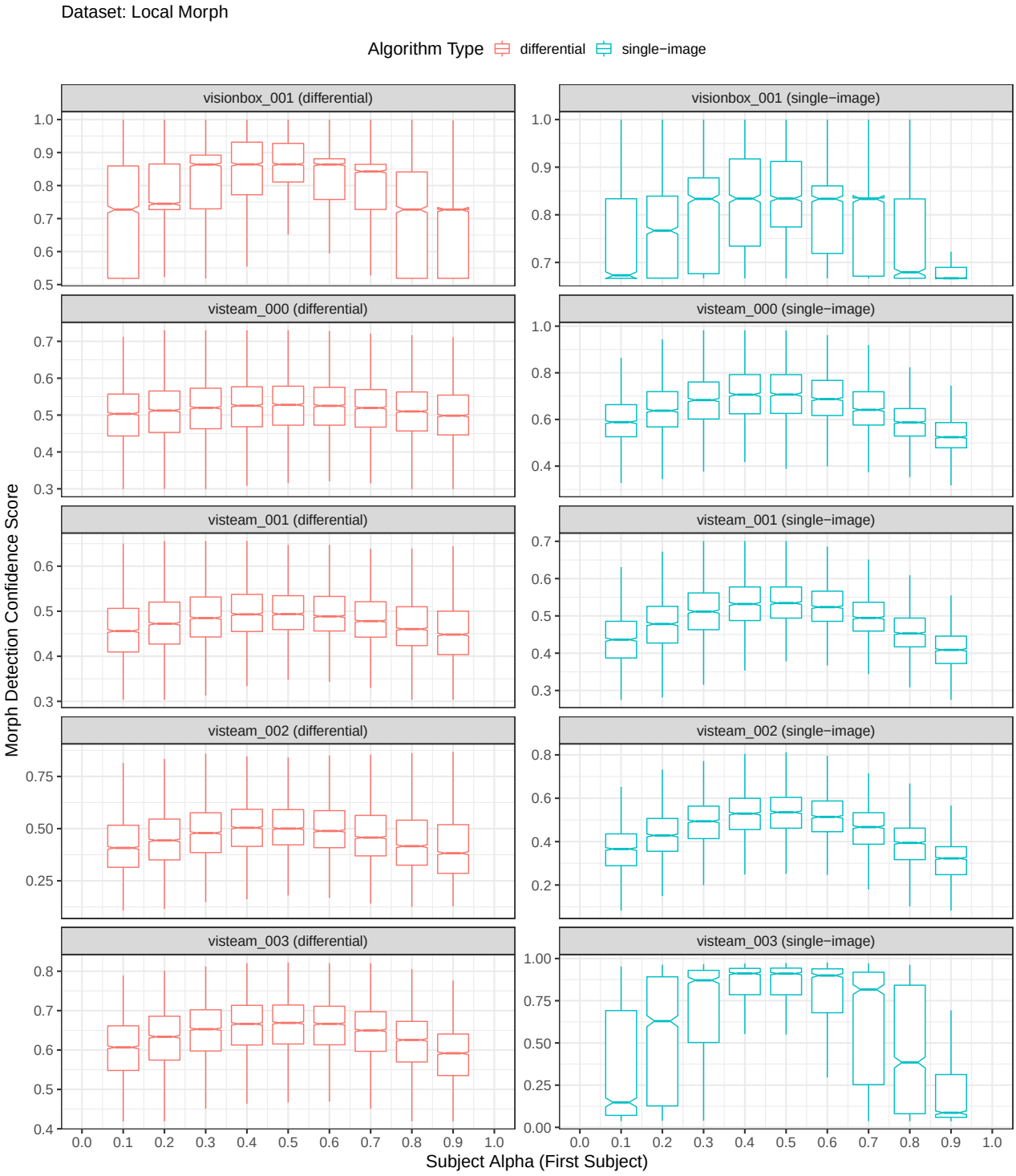


Figure 40: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

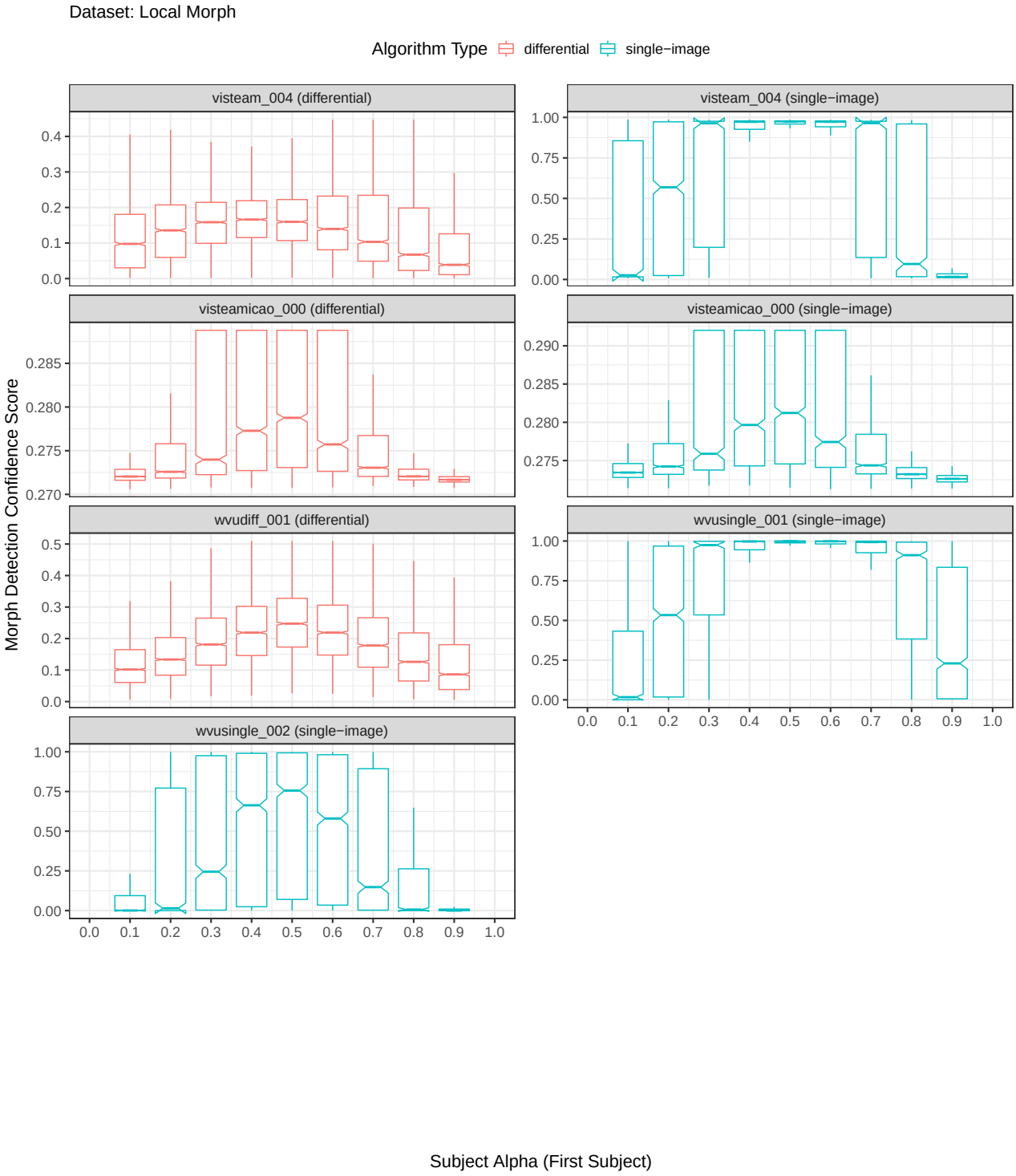


Figure 41: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

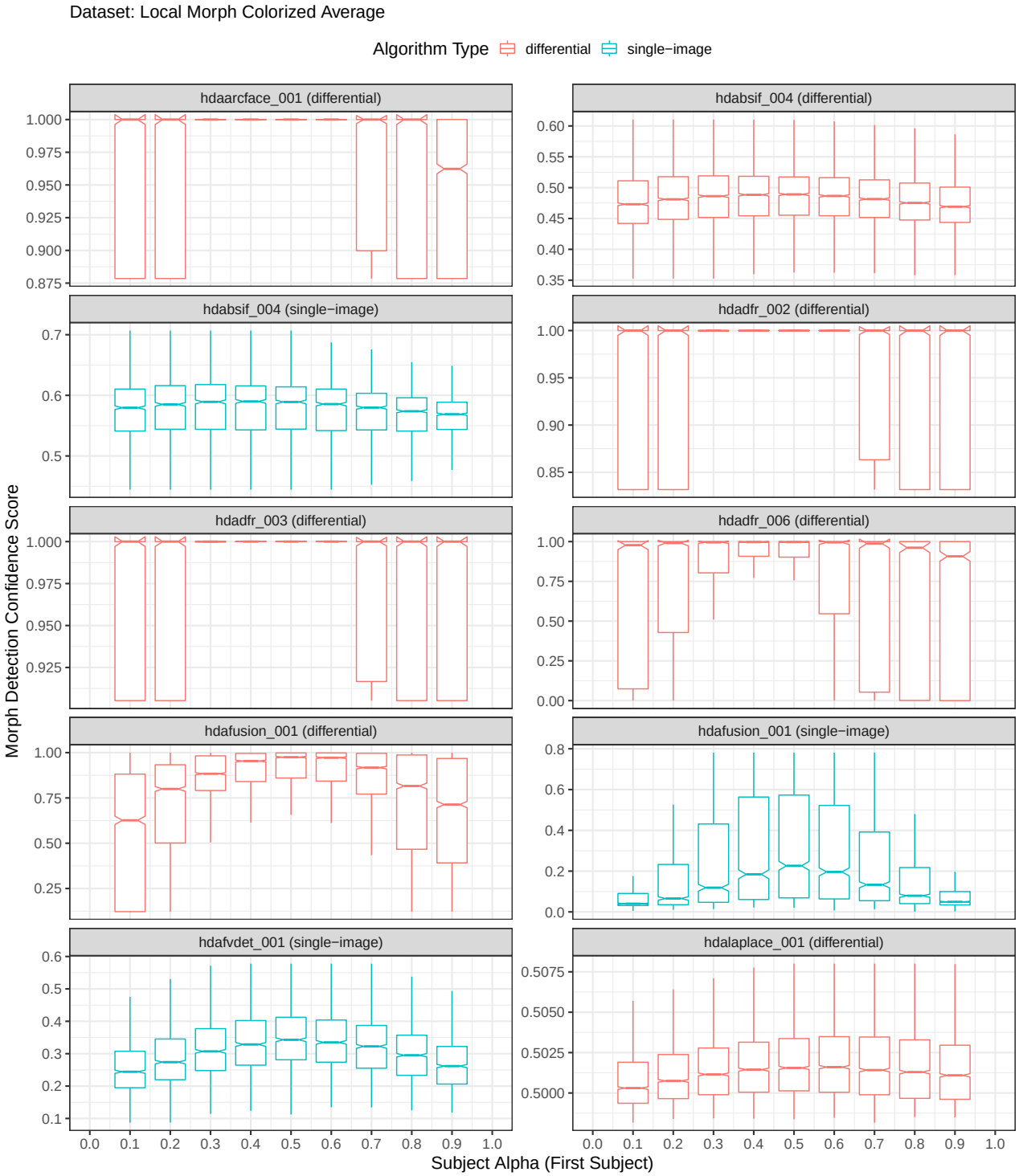


Figure 42: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

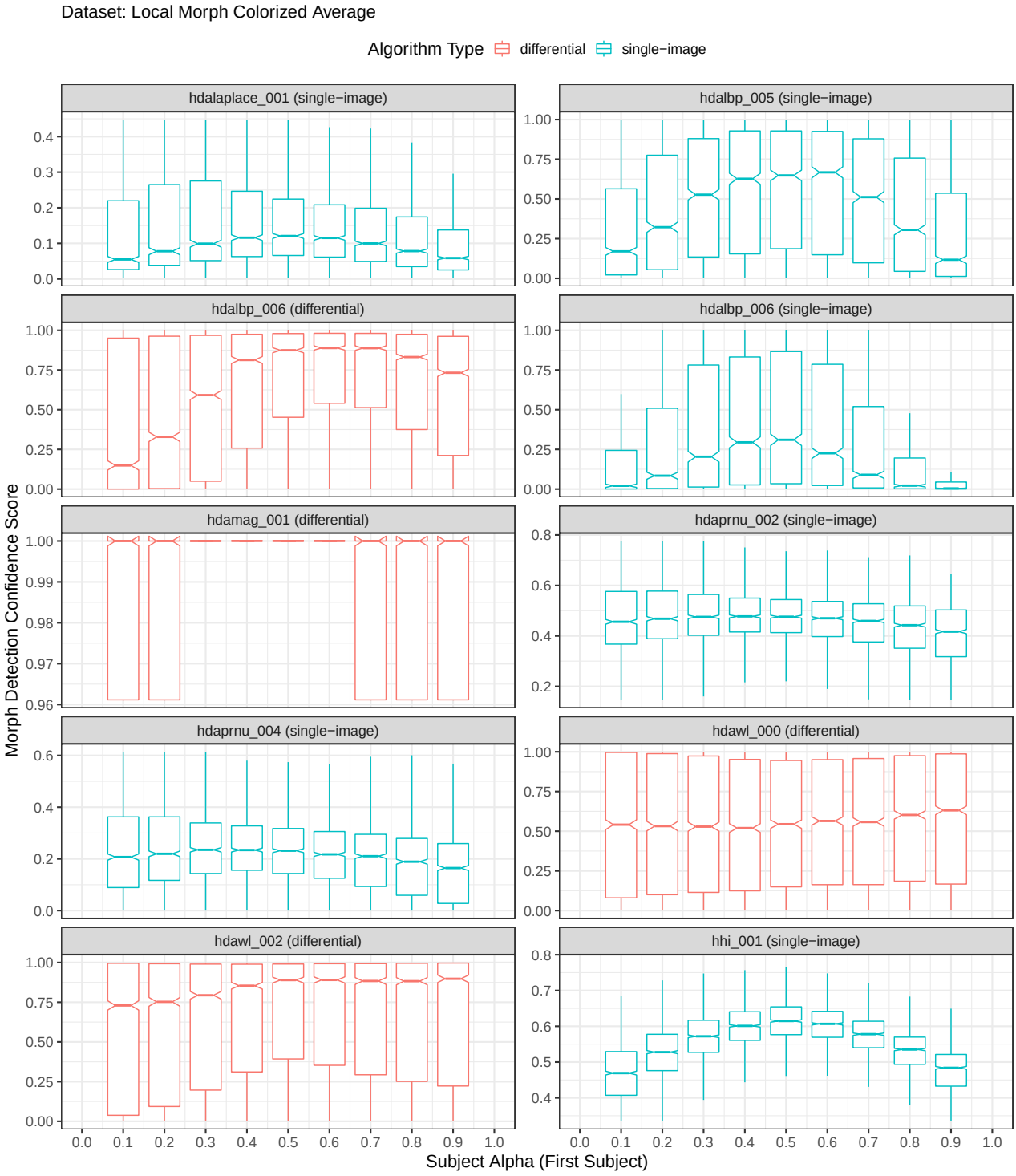


Figure 43: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

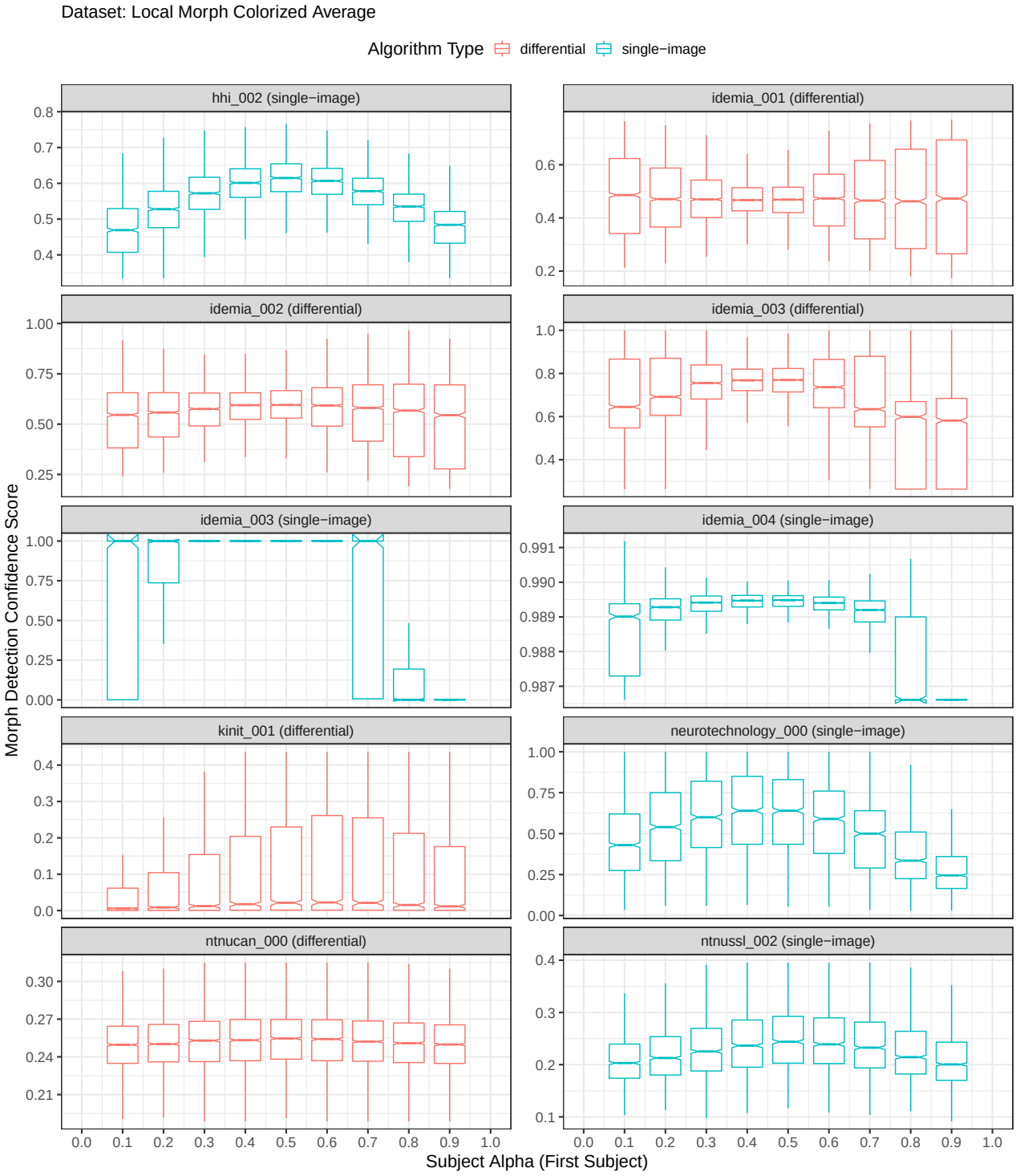


Figure 44: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

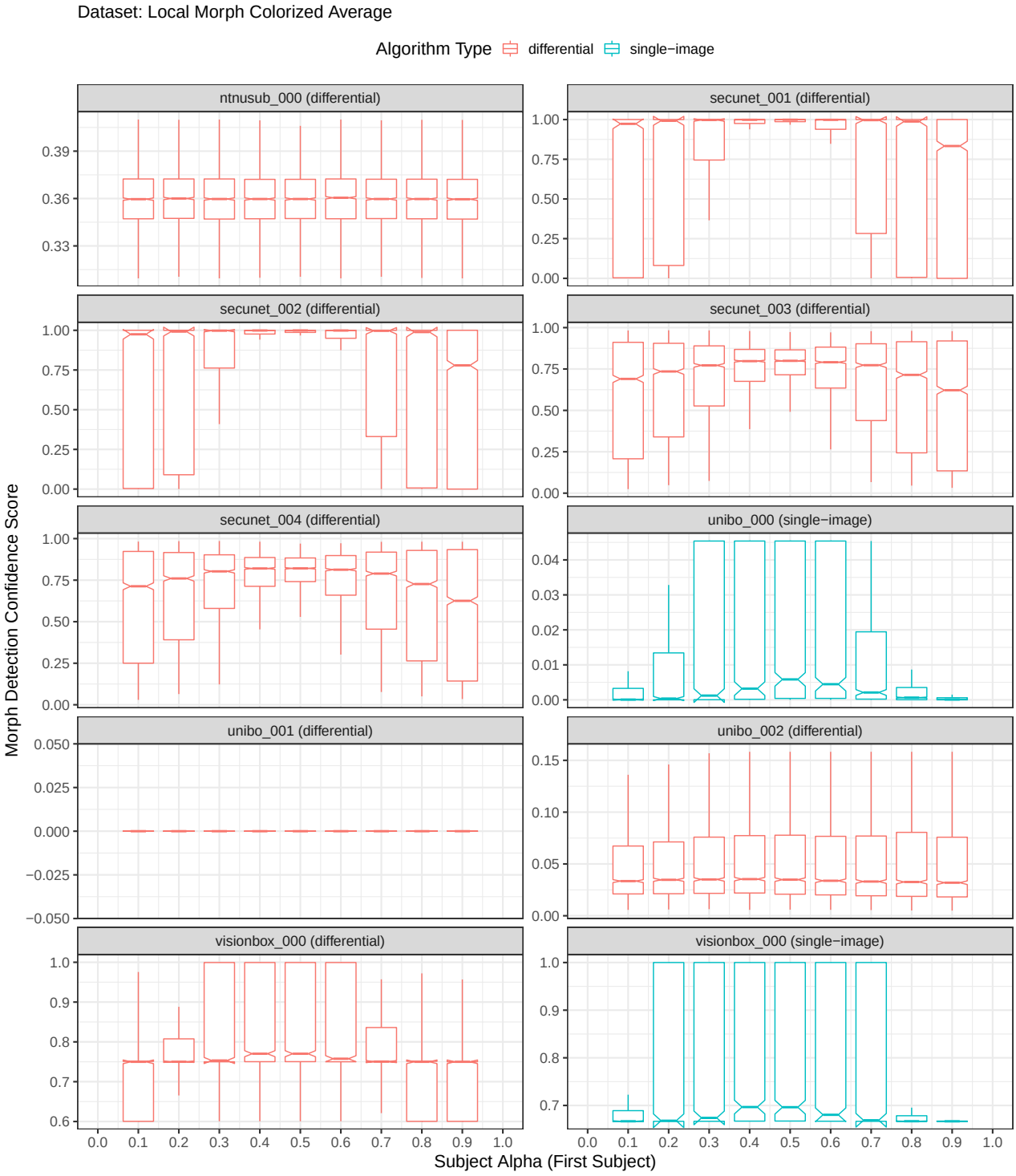


Figure 45: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

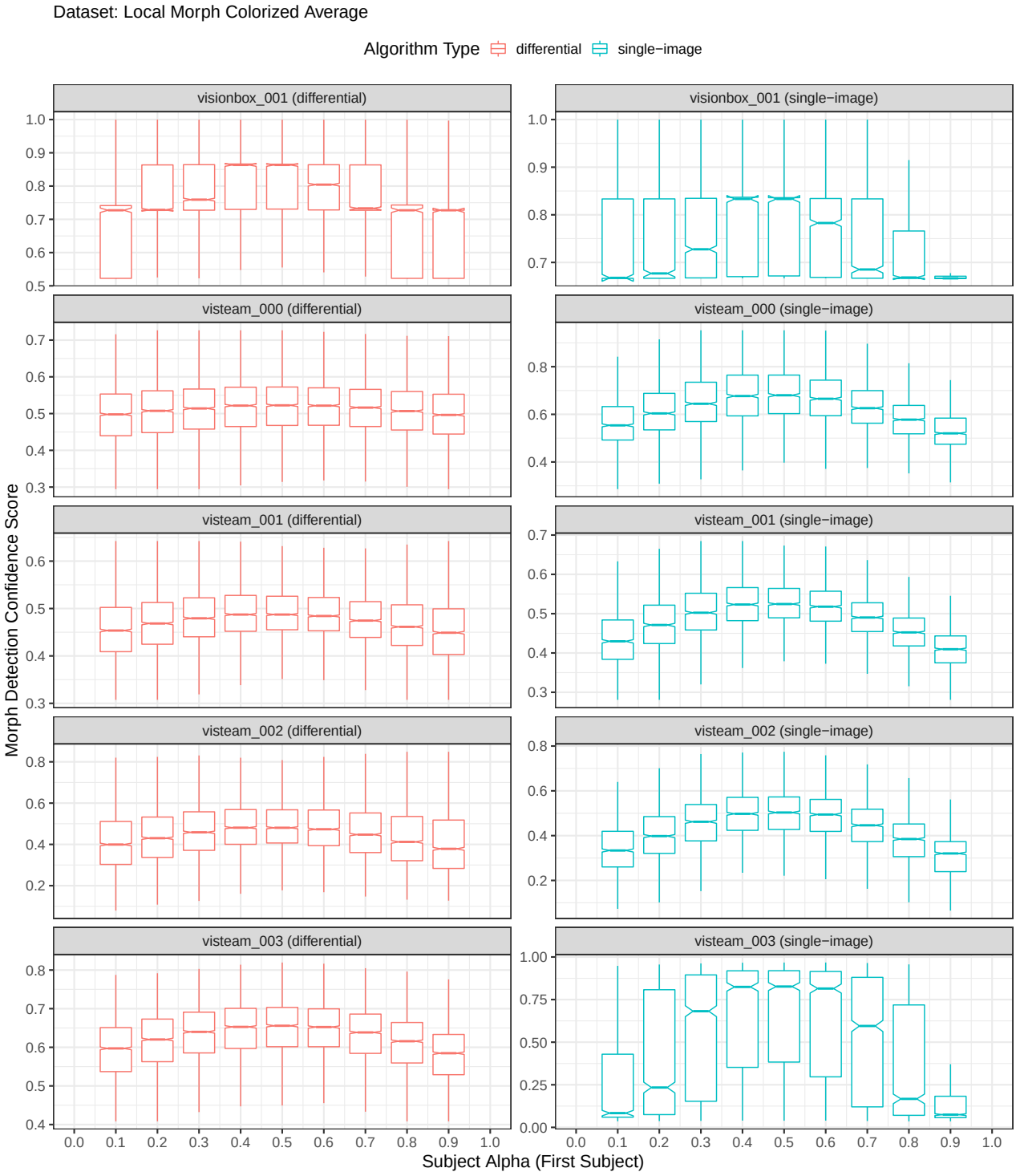


Figure 46: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

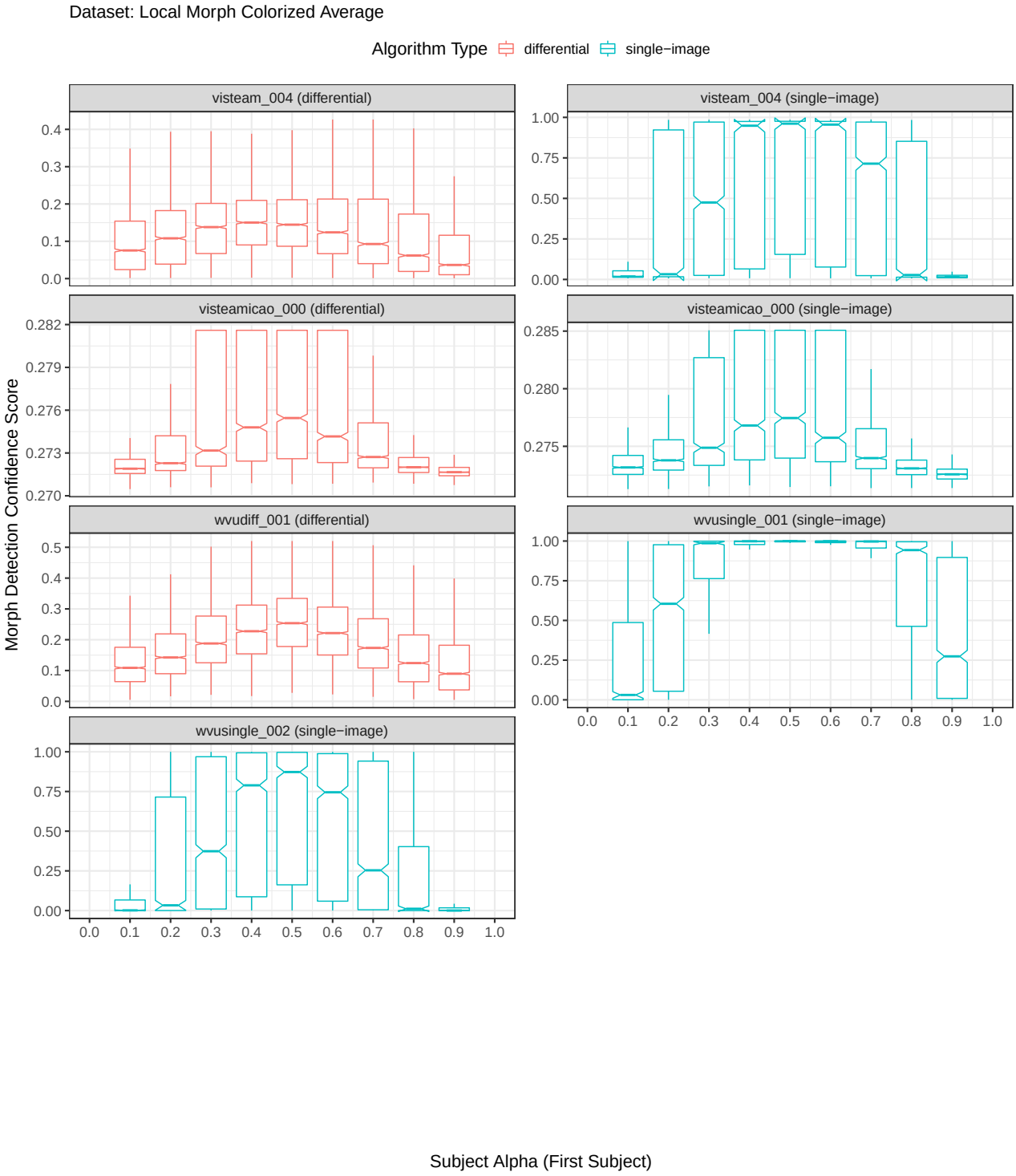


Figure 47: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

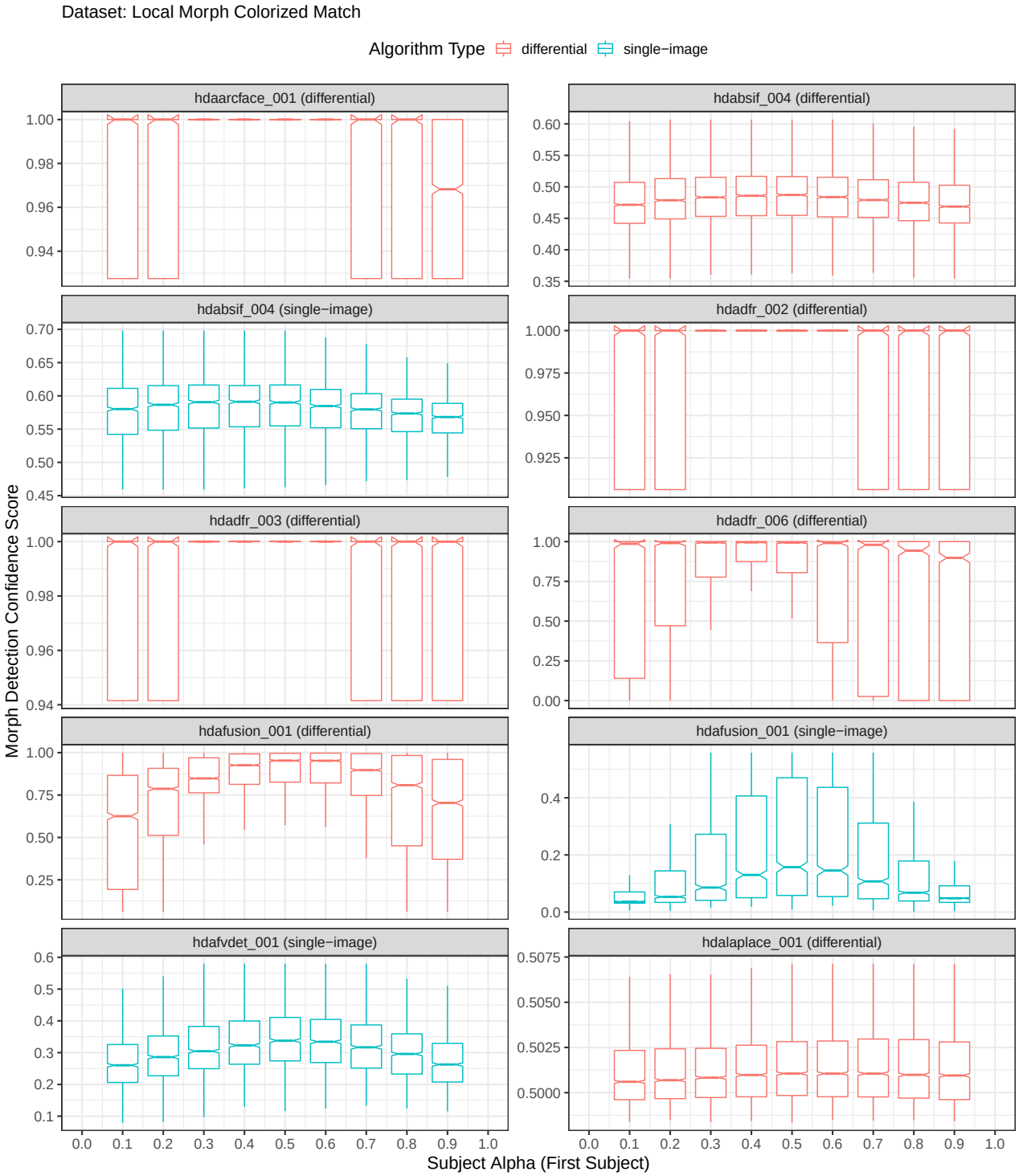


Figure 48: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

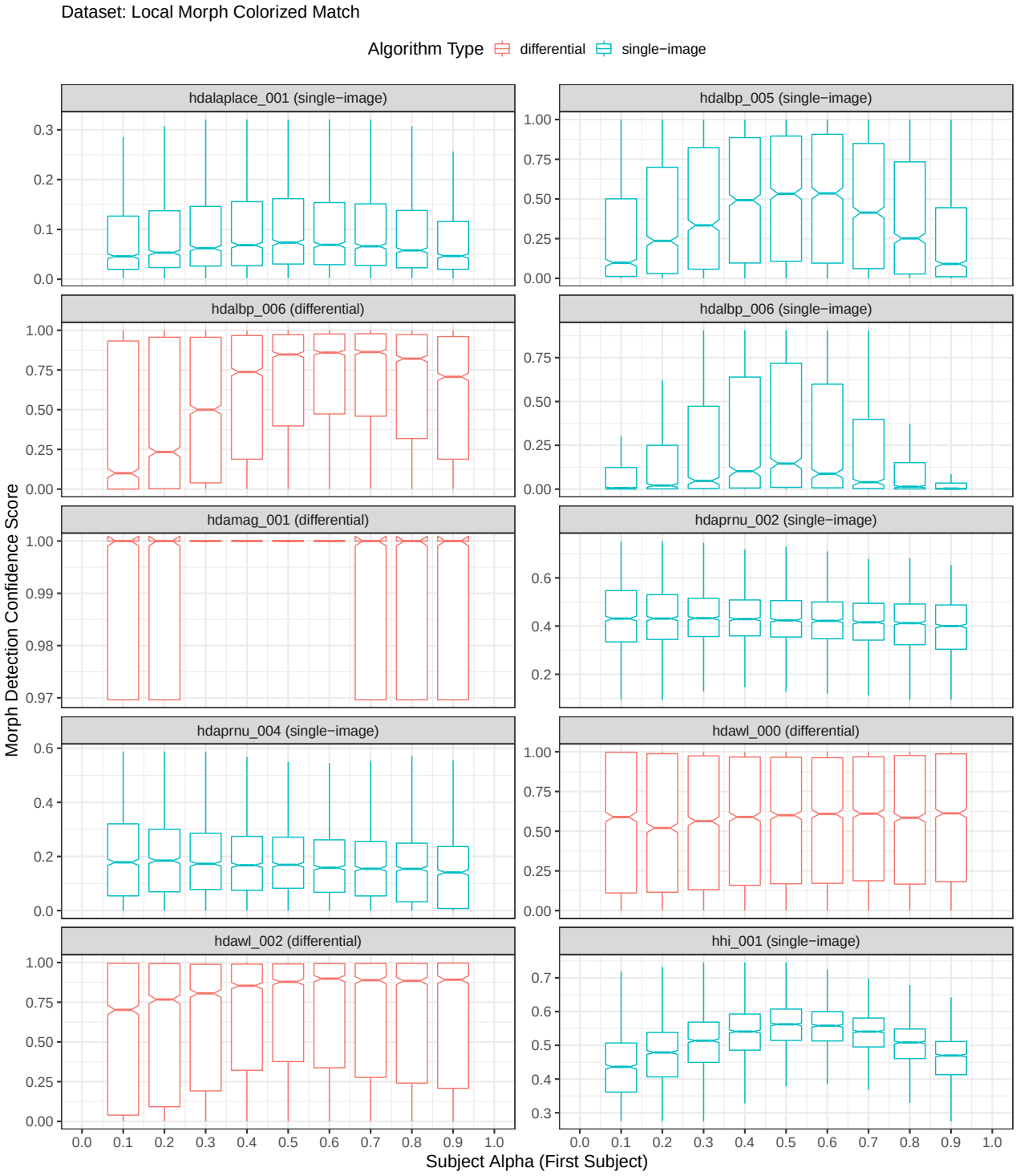


Figure 49: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

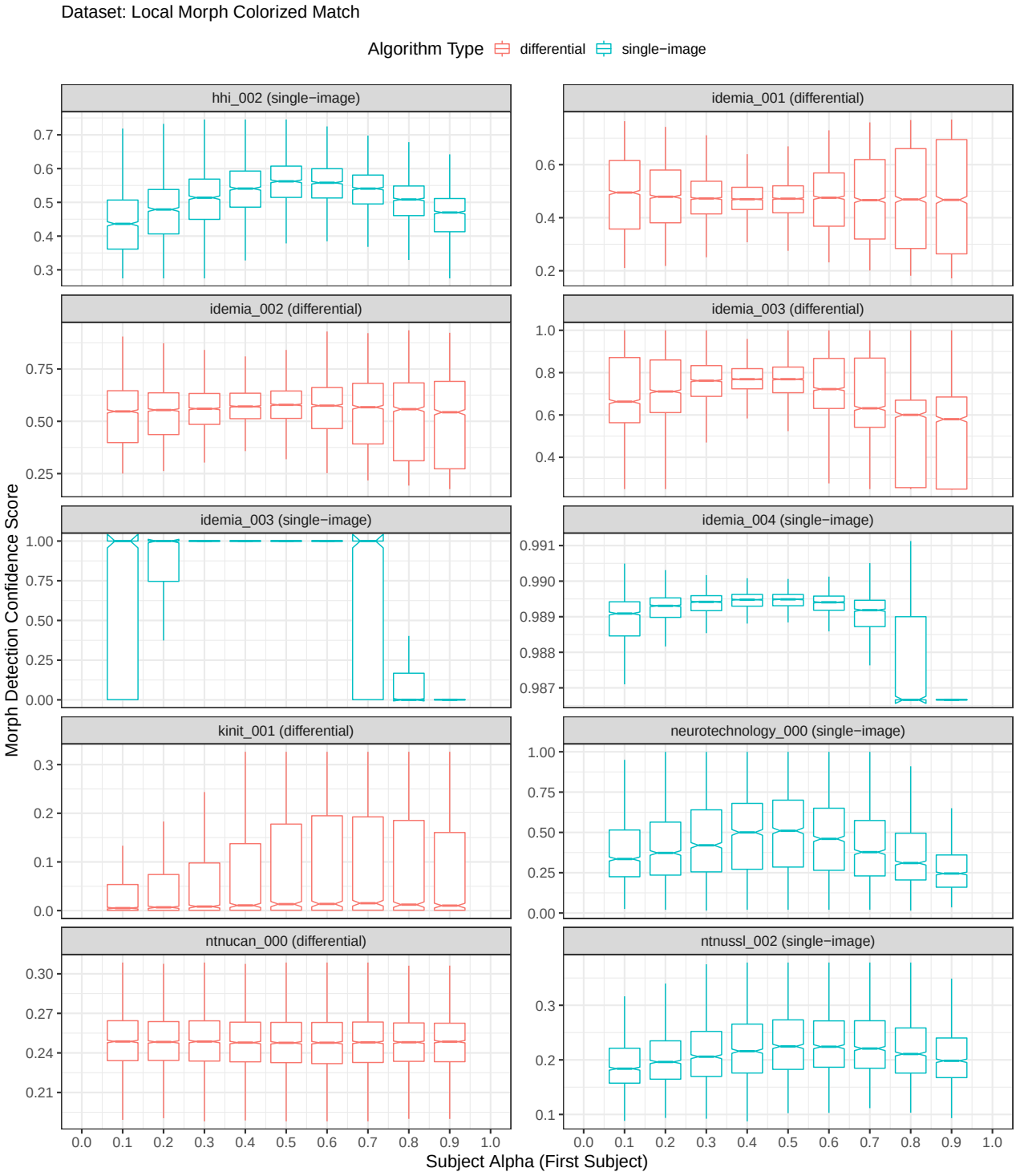


Figure 50: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

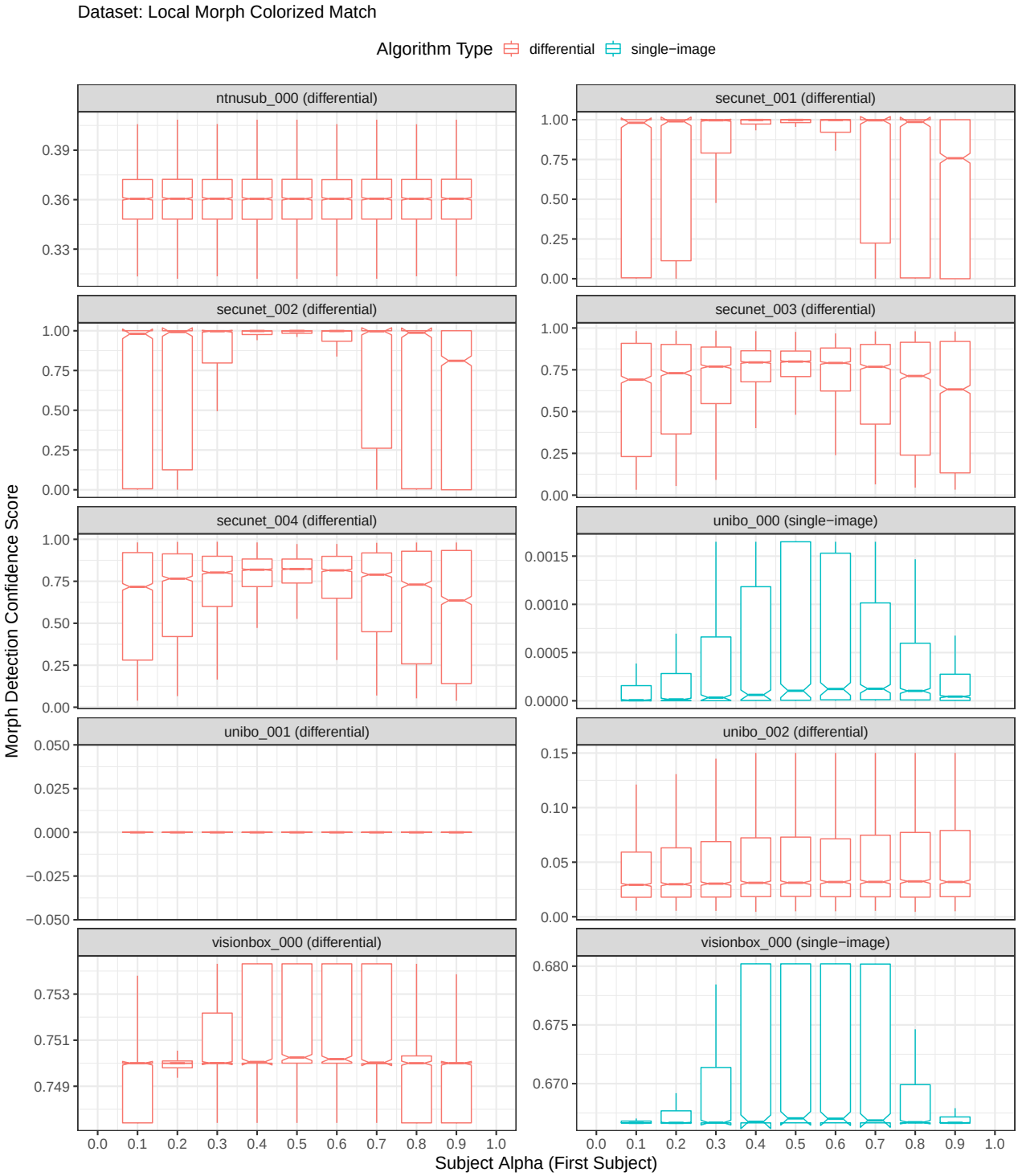


Figure 51: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

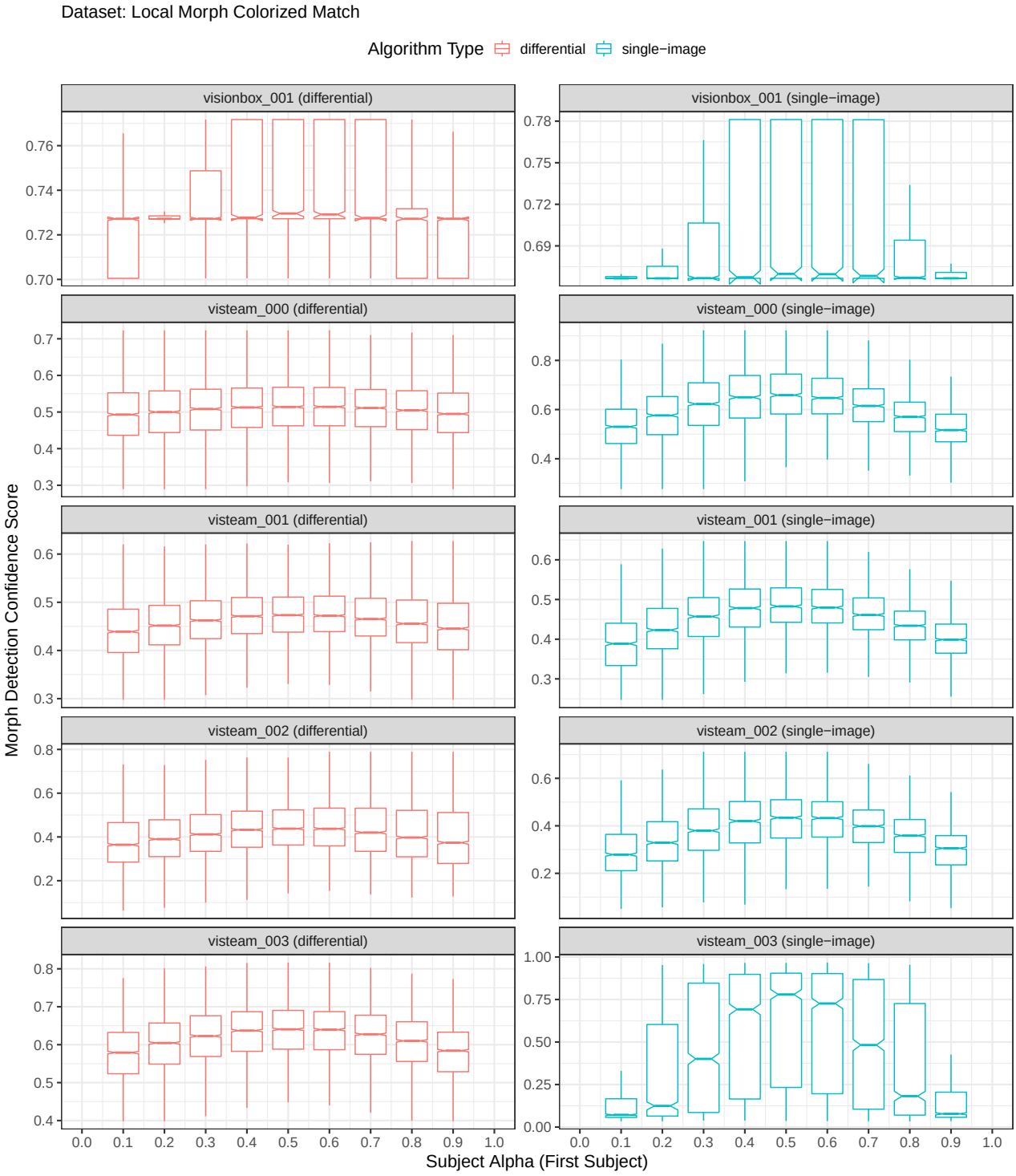


Figure 52: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

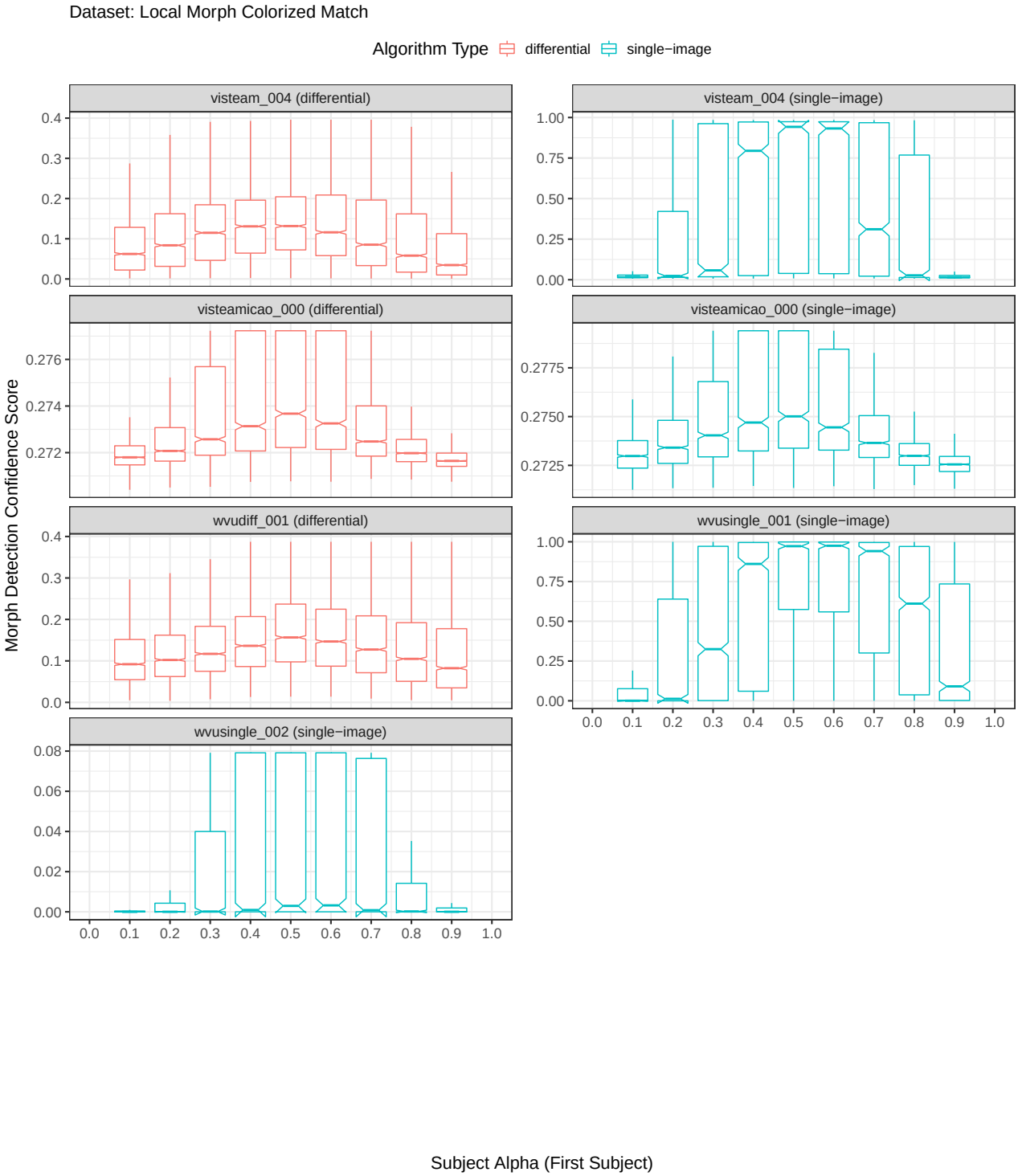


Figure 53: Boxplots plotting morph detection confidence score as a function of subject alpha (first subject in morph). Each plot includes scores that were successfully generated by the algorithm (i.e., results from failure to process were not used in this analysis). For individual algorithm results that are filterable and interactive, please refer to the algorithm report cards that are linked from the accuracy summary table on the FATE MORPH webpage.

4.11 Face Recognition Accuracy on Morphs

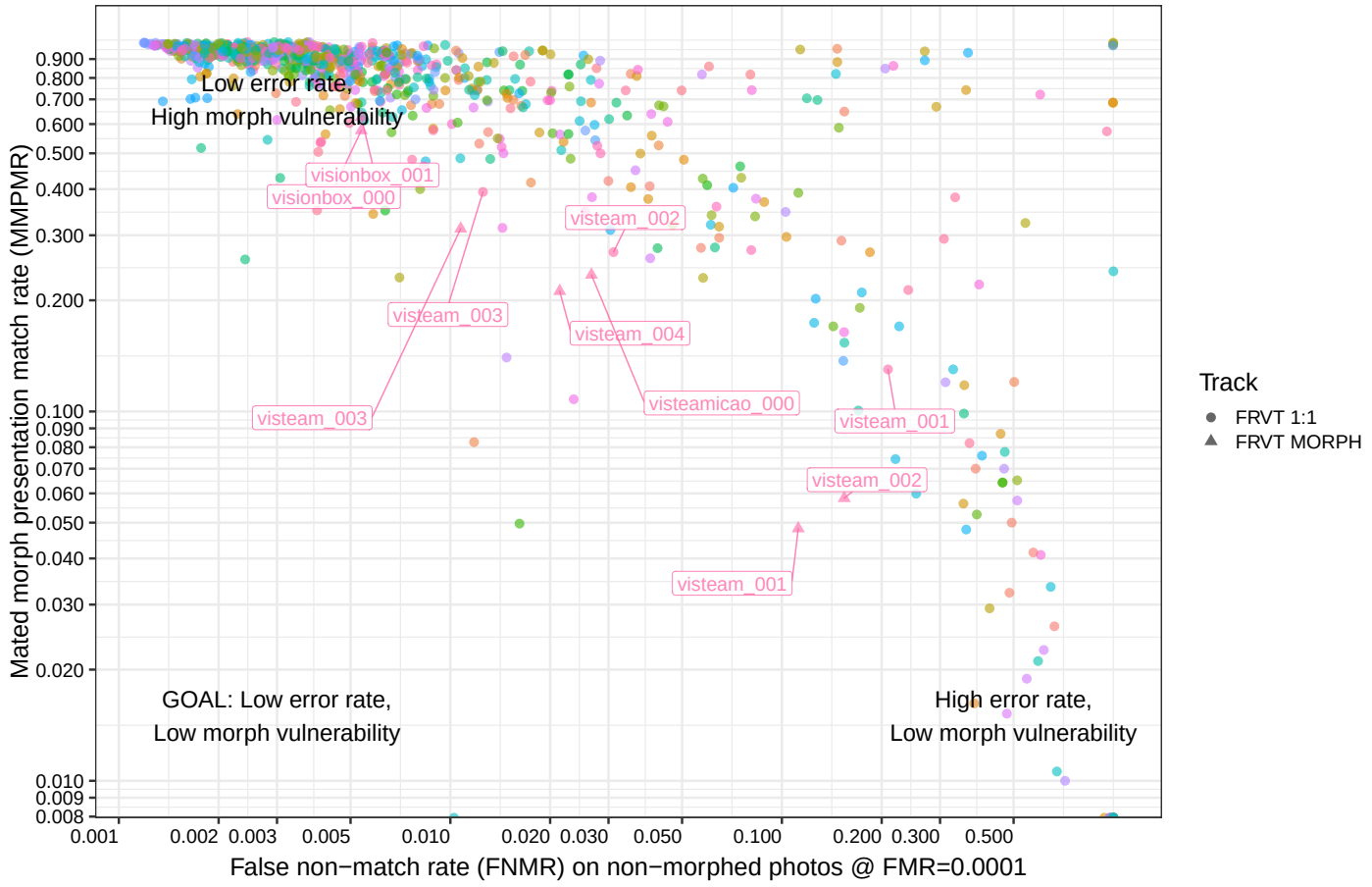


Figure 54: This graph plots face recognition algorithm vulnerability on morphs against general algorithm accuracy on non-morphed photos. Each circular point represents a face recognition algorithm recently submitted to the NIST FRTE 1:1 activity, and each triangular point represents a face recognition algorithm submitted to the NIST FATE MORPH activity. Note that algorithms submitted to FRTE 1:1 are not necessarily designed to handle morphed photos, and results are presented only as a point of reference. Submissions to FATE MORPH are designed with goals of face recognition algorithm resistance against morphing. The y-axis plots MMPMR, which is the fraction of morphs where both subjects incorrectly match to the morph. The x-axis plots FNMR or miss rate on regular photos, which provides an indication of general algorithm accuracy. Both MMPMR and FNMR are calculated with thresholds set to where the false match rate (FMR) is 0.0001. These results were generated with the Visa-Border morph dataset.

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[1] Matteo Ferrara, Annalisa Franco, and Davide Maltoni. *On the Effects of Image Alterations on Face Recognition Accuracy*, pages 195–222. Springer International Publishing, Cham, 2016.

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[3] M. Ferrara, A. Franco, and D. Maltoni. The magic passport. In *IEEE International Joint Conference on Biometrics*, pages 1–7, Sep. 2014.

[4] M. Ferrara, A. Franco, and D. Maltoni. Face demorphing. *IEEE Transactions on Information Forensics and Security*, 13(4):1008–1017, April 2018.

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[7] Ulrich Scherhag, Christian Rathgeb, Johannes Merkle, and Christoph Busch. Deep face representations for differential morphing attack detection, 2020.

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