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NISTIR 8331 DRAFT SUPPLEMENT

Ongoing Face Recognition Vendor Test (FRVT)

Part 6B: Face recognition accuracy with face masks using post-COVID-19 algorithms

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National Institute of Standards and Technology
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Executive Summary

OVERVIEW

This is a draft supplement to the second report on the performance of face recognition algorithms on faces occluded by protective face masks [2] commonly worn to reduce inhalation and exhalation of viruses. Inspired by the COVID-19 pandemic response, this is a continuous study being run under the Ongoing Face Recognition Vendor Test (FRVT) executed by the National Institute of Standards and Technology (NIST). In our first report [8], we tested "pre-pandemic" algorithms that were already submitted to FRVT 1:1 prior to mid-March 2020. This report augments its predecessor with results for more recent algorithms provided to NIST after mid-March 2020. While we do not have information on whether or not a particular algorithm was designed with face coverings in mind, the results show evidence that a number of developers have adapted their algorithms to support face recognition on subjects potentially wearing face masks. The algorithms tested were one-to-one algorithms submitted to the FRVT 1:1 Verification track.

WHAT'S NEW

This report includes

- ▷ Results from testing 100 face recognition algorithms provided to NIST since mid-March 2020
- ▷ Cumulative results for 198 algorithms evaluated to date (submitted both prior to and after mid-March 2020)

MOTIVATION

Traditionally, face recognition systems (in cooperative settings) are presented with mostly non-occluded faces, which include primary facial features such as the eyes, nose, and mouth. However, there are a number of circumstances in which faces are occluded by masks such as in pandemics, medical settings, excessive pollution, or laboratories. Inspired by the COVID-19 pandemic response, the widespread requirement that people wear protective face masks in public places has driven a need to understand how cooperative face recognition technology deals with occluded faces, often with just the periocular area and above visible. An increasing number of research publications have surfaced on the topic of face recognition on people wearing masks along with face-masked research datasets [10]. A number of commercial providers have announced the availability of face recognition algorithms capable of handling face masks, and this report documents performance results for 100 algorithms submitted to NIST after mid-March 2020. This report includes results for all algorithms evaluated to date. At the time of this writing, we are not aware of any large-scale, independent, and publicly reported evaluation on the effects of face mask occlusion on face recognition.

WHAT WE DID

The NIST Information Technology Laboratory (ITL) quantified the accuracy of face recognition algorithms on faces occluded by masks applied digitally to a large set of photos that has been used in an FRVT verification benchmark since 2018. These algorithms were submitted to FRVT 1:1 and includes 100 new algorithms provided to NIST since mid-March 2020. While we do not have information on whether or not a particular algorithm was designed with face coverings in mind, the algorithms were submitted with the expectation that NIST would execute them on masked face images. Using the original unmasked images to form a baseline for accuracy, we measured the impact of occlusion by digitally applying a mask to the face and varying mask shape, mask color, and nose coverage.

We ran these algorithms over a large set of photographs collected in U.S. governmental applications that are currently in operation: **application photographs** from a global population of applicants for immigration benefits and **border crossing photographs** of travelers entering the United States. Both datasets were collected for authorized travel or immigration processes.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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WHAT WE DID (CONTINUED)

The application photos (used as reference images) have good compliance with image capture standards. The border crossing photos (used as probe images) are not in good compliance with image capture standards given possible constraints on capture equipment, duration, facilities, and environment. We evaluated the case where the application photos were left unmasked, and synthetic masks were applied to the border crossing photos. This mimics an operational scenario where a person wearing a mask attempts to authenticate against a prior visa or passport photo. We also evaluated when both the application photos and border crossing photos were masked. This mimics, for example, a seamless travel scenario through an airport where a masked face image captured at check-in is enrolled and used during subsequent authentication attempts of the passenger still wearing a face mask. Together these datasets allowed us to process a total of 6.2 million images through a cumulative total of 198 algorithms.

Our use of software to apply masks to face images has the following advantages: it allows very rapid characterization of the effect of masks on face recognition; it allows controlled exploration of factors such as mask size, shape, and color; it affords repeatability, which is key to the fair comparison of algorithms; it scales to very large datasets - in our study, some 6.2 million photographs - which allows fine-grained characterization of false positive rates in addition to false negative rates. Conversely, our use of digital masks presents a number of limitations - please see the *Limitations* section of this executive summary for a more detailed discussion on the limitations of this study.

WHAT WE FOUND

The following results represent observations on algorithms provided to NIST both before and after the COVID-19 pandemic to date. We do not have information on whether or not a particular algorithm was designed with face coverings in mind. The results documented capture a snapshot of algorithms submitted to the FRVT 1:1 in face recognition on subjects potentially wearing face masks.

- ▷ **False rejection performance**: All algorithms submitted after the pandemic continue to give increased false non-match rates (FNMR) when the probes are masked. While a few pre-pandemic algorithms still remain within the most accurate on masked photos, a number of developers have submitted algorithms after the pandemic showing significantly improved accuracy and are now among the most accurate in our test. Using border crossing images, without masks, the most accurate algorithms will fail to authenticate about 0.3% of persons while falsely accepting no more than 1 in 100000 impostors (i.e. FNMR$= 0.003$ at FMR$= 0.00001$). With the highest coverage mask we tested and the most accurate algorithms, this failure rate rises to about 3% (FNMR = 0.03) among all algorithms. This is noteworthy given that around 70% of the face area is occluded by the mask. However, many algorithms submitted since mid-March 2020 remain much less tolerant: some algorithms that are quite competitive with unmasked faces (FNMR < 0.01) still fail to authenticate between 10% to 40% of masked images (FNMR $\rightarrow 0.4$).

See Figures 10, 11 and Table 6

For the case where both the enrollment and verification images are masked, interestingly, many algorithms show a reduction in false non-match rates compared to when only the verification image is masked, at a fixed threshold. While the reduction in FNMR is favorable, we observe much larger false match rates when both images are masked. These findings are discussed in subsequent sections of this executive summary. *See Figure 49*

In cooperative access control applications, false rejections can traditionally be remedied by users making second attempts. This is effective when users correct pose, expression, or illumination aspects of their presentation. With masked faces, however, a second attempt may not be effective if the failure is a systematic property of the algorithm.

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WHAT WE FOUND (CONTINUED)

▷ **Evolution of algorithms on face masks**: We observe that a number of algorithms submitted since mid-March 2020 show notable reductions in error rates with face masks over their pre-pandemic predecessors. When comparing error rates for unmasked versus masked faces, the median FNMR across algorithms submitted since mid-March 2020 has been reduced by around 25% from the median pre-pandemic results. The figure below presents examples of developer evolution on both masked and unmasked datasets. For some developers, false rejection rates in their algorithms submitted since mid-March 2020 decreased by as much as a factor of 10 over their pre-pandemic algorithms, which is evidence that some providers are adapting their algorithms to handle face masks. However, in the best cases, when comparing results for unmasked images to masked images, false rejection rates have increased from 0.3%-0.5% (unmasked) to 1.6%-5% (masked). The current performance of face recognition with face masks is comparable to the state-of-the-art on unmasked images in mid-2017 [4].

Figure 1: Examples of developer evolution of accuracy on masked and unmasked datasets.

See Figures 10, 5, and 11

▷ **False acceptance performance**: As most systems are configured with a fixed threshold, it is necessary to report both false negative and false positive rates for each group at that threshold. When comparing a masked probe to an unmasked enrollment photo, in most cases, false match rates (FMR) are reduced by masks. The effect is generally modest with reductions in FMR usually being smaller than a factor of two. This property is valuable in that masked probes do not impart adverse false match security consequences for verification.

However, when both the enrollment and verification images are masked, most algorithms give elevated false match rates, with FMR ranging from 10 to 100 times higher than when only the probe is masked or both images are unmasked, at the same threshold. This behavior applies to most algorithms tested, with the exception of particular algorithms from a small number of developers (e.g., idemia-009, pensees-001, fujitsulab-002, neurotechnology-010, glory-003). *See Figure 49*

▷ **Mask-agnostic face recognition**: All 1:1 verification algorithms submitted to the FRVT test since the start of the pandemic are evaluated on both masked and unmasked datasets. The test is designed this way to mimic operational reality: some images will have masks, some will not (especially enrollment samples from a database or ID card). And to the extent that the use of protective masks will exist for some time, our test will continue to evaluate algorithmic capability on verifying all combinations of masked and unmasked faces.

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WHAT WE FOUND (CONTINUED)

At least three developers - Idemia, Pensees, and Neurotechnology - have developed algorithms that work with any combination of masked and unmasked images, generating approximately constant FMR across any masked/unmasked combination, and similarly, yield approximately constant FNMR across masked-probe and masked-enrollment-and-probe combinations.

Figure 2: FNMR and FMR calibration curves on masked/unmasked combinations.

An example of an integrated approach might be: 1) inclusion of a mask detector 2) on an unmasked image, extract features from both the full face and the periocular region 3) on a detected masked image, extract features from the periocular region 4) at match time, compare full-face templates when both images are unmasked, and periocular templates otherwise.

- **Coverage of the masks**: Masks that occlude more of the face give larger false non-match rates. We surveyed over the extent to which the mask covers the nose, from not at all ("low") to typical ("medium") to near the eyes ("high"). We baselined those with unmasked faces with the result that FNMR increases by factors of around 5, 18, and 24 respectively for the median algorithm submitted since mid-March 2020. However, as noted, algorithms vary considerably in the their tolerance of coverage. Readers should consult tabulated values for specific algorithms.

  See Table 6 and Figures 13, 15

  We included the "low" option not because it is a common position for a mask but as an option for authentication applications where it would be tenable to ask the user to pull the mask down to just below the nose for the duration of the authentication attempt.

- **Color of the masks**: We considered white, light-blue, red, and black masks. Many algorithms have higher error rates in black and red masks than light-blue and white masks. The reason for observed accuracy differences between mask color is unknown but is a point for consideration by impacted developers. Mask color also affects the rate at which some algorithms fail to produce a template from an image. *See Figure 32 and Table 11*

- **Shape of the masks**: The shape of the masks matters. Full-face-width masks generally cover more of the face than rounder N95 type masks. Post-pandemic algorithm results show that wide-width masks generally give false negative rates about a factor of 1.5 higher than do rounder type masks. *See Figure 12*

- **Failure to detect and template:** The false negative rates in this report include the effects of both face detection and localization errors, and low-similarity matching errors. We separately include tables detailing how often an algorithm does not make a template from an input image. While many algorithms give low failure-to-template rates, some give high values ranging close to 100%. Conversely, the successful creation of a template does not guarantee proper facial localization. Such localization failures will not be captured as a failure to detect and template event but will impact accuracy rates nonetheless. *See Table 11 and Figure 14*

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LIMITATIONS As a simulation, this study likely doesn't fully capture the effects of masks on face recognition. Particularly the following points should be weighed by readers in the near term. Some of these will be addressed in subsequent work at NIST.

- **Train algorithms:** As with all NIST evaluations, we regard the software as a black box whose parameters (models) remain fixed for the entirety of its use without learning from the test data. We do not train or fine-tune algorithms.
- **Evaluate one-to-many algorithms:** We have only run one-to-one verification algorithms with masks. This elicits data on the effect of masks on the underlying feature extraction and discrimination of algorithms and can therefore be be expected to give first-order indications of the effect on one-to-many identification algorithms.
- **Consider the effect of eye occlusion:** We did not address the effect of eye-glasses or eye-protection. While our dataset includes examples of people wearing glasses, we didn't collect such data nor simulate it with digital addition.
- **Test with images of real masks:** Given time and resource constraints, we didn't collect photos of subjects wearing masks. The possible downsides of this are several. First, our digital masks are tailored to faces; while a few don't fit realistically, mass-produced real masks may not fit all actual persons correctly either. We were not able to pursue an exhaustive simulation of the endless variations in color, design, shape, texture, bands, and ways masks can be worn. Second, because many cameras run with exposure-control, it is possible that a dark mask will cause less light to be reflecting and the camera to increase gain on the sensor causing overexposure of the periocular region. Likewise a white mask could lead to underexposure problems. Third, it is possible that some cameras that include a face detector, may fail to focus or acquire a masked face correctly.
- **Use textured masks:** All masks synthesized by NIST in this study have a uniform color. The consequences of this are that we do not capture the impact of mask texture or pattern on face recognition. The possibility exists for patterned masks to induce higher facial localization errors, which is not captured in our current study. We received a suggestion that such information may serve as a soft biometric, in that a subject that always wears the same textured mask will be more identifiable. We don't intend to encourage algorithm development along this line, because as mass-produced high-efficacy masks become more common, mask diversity may actually drop.
- **Study demographic effects on masked images:** This report estimates overall performance of existing algorithms on recognition of faces occluded by masks. We deferred tabulating accuracy for different demographic groups until more capable mask-enabled algorithms have been submitted to FRVT.
- **Evaluate algorithms on non-cooperative, unconstrained imagery:** This report documents results for matching masked webcam images to unmasked portrait-style photos. While the properties of the two sets of images differ, subjects are operating in cooperative mode and are for the most part, looking at the camera.
- **Consider effects of human examination:** This report does not consider the various ways humans are involved in face recognition systems. For example, analysts can correct face detection or localization errors induced by masks, prior to automated recognition. Likewise, humans are often tasked with adjudication of images following a rejection or other exception from an automated system. Analysis of human capability and role is pertinent to those operations, but is beyond the scope of this study.

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IMPLICATIONS AND FUTURE WORK

Know Your Algorithm: Operational implementations usually employ a single face recognition algorithm. Given algorithm-specific sensitivities to masks and other image or subject properties, it is incumbent upon the system owner to know their algorithm. While publicly available test data from NIST and elsewhere can inform owners, it will usually be informative to specifically measure accuracy of the operational algorithm on the operational image data collected with actual masks.

NIST plans on releasing a series of reports, iteratively assessing different aspects and use cases of face masking on recognition performance.

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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.

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 Face Mask Effects

1.1 Status

NIST has conducted a series of tests aimed at quantifying face recognition accuracy for people wearing masks. Our initial approach has been to apply masks to faces digitally (i.e., using software to apply a synthetic mask). This allowed us to leverage large datasets that we already have. This report documents results for 1:1 verification algorithms. In our first report [8], we tested "pre-pandemic" algorithms that were already submitted to FRVT 1:1 prior to mid-March 2020. This report augments its predecessor with results for more recent algorithms provided to NIST **after the COVID-19 pandemic**. While we do not have information on whether or not a particular algorithm was designed with face coverings in mind, the algorithms were submitted with the expectation that NIST would execute them on masked face images. In addition to reporting results for when only the verification image is masked, we also document the effects for the case when both enrollment and verification images are masked. This report quantifies the effect of masks on both false negative and false positives match rates and tracks developer evolution of face recognition accuracy with face masks.

The FRVT evaluation is an ongoing test that remains open to new participation. Comments and suggestions should be directed to frvt@nist.gov.

1.2 Introduction

The majority of face recognition systems have been deployed in applications where subjects make cooperative presentations to a camera, for example as part of an application for a benefit or ID credential, or as during access control. With very few exceptions such images would not include face masks or other occlusions. However, with the COVID-19 pandemic, we can anticipate a demand to authenticate persons wearing masks, for example in immigration settings, without the need to the subjects to remove those masks. This presents a problem for face recognition, because regions of the face occluded by masks - the mouth and nose - include information useful for both recognition and, potentially, the detection stage that precedes it.

Previous work on face recognition of occluded faces has been directed at situations such as crime scenes where subjects were actively un-cooperative i.e. acting to evade face detection and recognition. Those applications are often characterized by image properties (low resolution, video compression, uncontrolled head orientation) that are known [5] to degrade recognition accuracy.

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2 Image Datasets

2.1 Application Images

The images are collected in an attended interview setting using dedicated capture equipment and lighting. The images are of size 300x300 pixels. The images are all high-quality frontal portraits collected in immigration offices and with a white background. As such, potential quality related drivers of high false match rates (such as blur) can be expected to be absent. The images are encoded as ISO/IEC 10918 i.e. JPEG. Over a random sample of 1000 images, the images have compressed file sizes (mean: 42KB, median: 58KB, 25-th percentile: 15KB, and 75-th percentile: 66KB). The implied bit-rates are mostly benign and superior to many e-Passports. When these images are provided as input into the algorithm, they are labeled with the type "ISO".

Figure 3: Examples of images with properties similar to the enrollment application photos used in this study. The subjects in the photos are all NIST employees.

2.2 Webcam Images

These images are taken with a camera oriented by an attendant toward a cooperating subject. This is done under time constraints, so there are roll, pitch, and yaw angle variations. Also, background illumination is sometimes bright, so the face is under exposed. Sometimes, there is perspective distortion due to close range images. The images are generally in poor conformance with the ISO/IEC 19794-5 Full Frontal image type. The images have mean interocular distance of 38 pixels. The images are all live capture. When these images are provided as input into the algorithm, they are labeled with the type "WILD". Examples of such images are included in Figure 4 and Figure 4 in NIST Interagency Report 8271. Results for verification of these images (unmasked) appear in FRVT Part 1 - Verification both compared against images of the same type, and with those described in section 2.1.

Description	#
Total images	6 244 865
Application (enrollment) images	1 019 232
Subjects in application images	1 019 232
Webcam (verification) images	5 225 633
Subjects in webcam (verification) images	2 535 329
Mated comparisons	3 225 633
Impostor comparisons	200 000 000
Subjects in mated comparisons	535 329
Subjects in impostor comparisons	3 019 232

Table 1: Summary quantities of the dataset used in this evaluation.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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2.3 Synthetically Masked Images

In this test, synthetically-generated masks were overlaid on top of 1) just the probe image (webcam images described in Section 2.2) or 2) both the enrollment (application photos described in Section 2.1) and probe images. The Dlib [7] C++ toolkit version 19.19 was used to detect and establish key facial points on the face, and with the facial points, solid masks of different shape, height, and color were drawn on the face. The exact Dlib facial points and details used to generate the masks are documented in Appendix A. In the event that Dlib was unable to detect a face in the image, eye coordinates were used to generate a mask leveraging standardized token frontal geometry [1].

Examples of synthetically-masked probe images are presented in Figures 4.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 4: Examples of synthetically-generated face masks used in this study. The original images are from the NIST MEDS-II Dataset [3]. They were collected in operational settings using the same camera and procedure as is used for the border images that form the mainstay of the experiments in this report.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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3 Metrics

3.1 Matching accuracy

Given a vector of N genuine scores, u, the false non-match rate (FNMR) is computed as the proportion below some threshold, T:

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

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 (FMR) is computed as the proportion above T:

$$\text{FMR}(T) = \frac{1}{N}\sum_{i=1}^{N} H(v_i - T) \tag{2}$$

The threshold, T, can take on any value. We typically generate a set of thresholds from quantiles of the observed impostor scores, v, as follows. Given some interesting false match rate range, $[\text{FMR}_L, \text{FMR}_U]$, we form a vector of K thresholds corresponding to FMR measurements evenly spaced on a logarithmic scale

$$T_k = Q(1 - \text{FMR}_k) \tag{3}$$

where Q is the quantile function, and FMR_k comes from

$$\log_{10}\text{FMR}_k = \log_{10}\text{FMR}_L + \frac{k}{K}\,[\log_{10}\text{FMR}_U - \log_{10}\text{FMR}_L] \tag{4}$$

Detection error tradeoff (DET) characteristics are plots of FNMR(T) vs. FMR(T). These are plotted with $\text{FMR}_U \to 1$ and FMR_L as low as is sustained by the number of impostor comparisons, N. This is somewhat higher than the "rule of three" limit $3/N$ [6] because samples are not independent, due to re-use of images.

3.2 Failure to Enroll

Failure to enroll (FTE) is the proportion of failed template generation attempts. Failures can occur because the software throws an exception, or because the software electively refuses to process the input image. This would typically occur if a face is not detected. FTE is measured as the number of function calls that give EITHER a non-zero error code OR that give a "small" template. This is defined as one whose size is less than 60 bytes. This second rule is needed because some algorithms incorrectly fail to return a non-zero error code when template generation fails yet do return a valid default data structure.

The effects of FTE are included in the accuracy results of this report by regarding any template comparison involving a failed template to produce a low similarity score. Thus higher FTE results in higher FNMR and lower FMR.

4 Algorithms

The FRVT activity is open to participation worldwide, and the test will evaluate submissions on an ongoing basis. There is no charge to participate. The process and format of algorithm submissions to NIST are described in the FRVT 1:1 Verification Application Programming Interface (API) [9] document. Participants provide their submissions in the form of libraries compiled on a specific 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.

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This report documents the results of algorithms submitted to FRVT 1:1 for testing both before (prior to mid-March 2020) and after the COVID-19 pandemic. Table 5 lists the algorithms that were tested. Note that algorithms that are expired or retired are not included in this report.

	Developer	Algorithm	Submission Date
1	3Divi	3divi-004	2019-07-22
2	3Divi	3divi-005	2020-08-28
3	ACI Software	acisw-003	2020-08-03
4	ACI Software	acisw-006	2021-02-25
5	ADVANCE.AI	advance-002	2019-12-19
6	ASUSTek Computer Inc	asusaics-000	2019-10-24
7	AYF Technology	ayftech-001	2020-07-06
8	Ability Enterprise - Andro Video	androvideo-000	2021-01-25
9	Acer Incorporated	acer-000	2020-01-08
10	Acer Incorporated	acer-001	2020-06-30
11	Adera Global PTE	adera-002	2021-02-16
12	Ai First	aifirst-001	2019-11-21
13	AiUnion Technology	aiunionface-000	2019-10-22
14	Aigen	aigen-001	2020-10-06
15	Akurat Satu Indonesia	ptakuratsatu-000	2020-09-11
16	AlphaSSTG	alphaface-002	2020-02-20
17	Anke Investments	anke-005	2019-11-21
18	Antheus Technologia	antheus-000	2019-12-05
19	Antheus Technologia	antheus-001	2020-06-25
20	AnyVision	anyvision-005	2021-02-03
21	Aware	aware-005	2020-02-27
22	Awidit Systems	awiros-001	2019-09-23
23	Awidit Systems	awiros-002	2020-10-28
24	Beihang University-ERCACAT	ercacat-001	2020-07-06
25	Beijing Alleyes Technology	alleyes-000	2020-03-09
26	BioID Technologies SA	bioidtechswiss-001	2020-08-28
27	BioID Technologies SA	bioidtechswiss-002	2021-02-17
28	Bresee Technology	bresee-000	2020-08-07
29	Bresee Technology	bresee-001	2020-12-30
30	CSA IntelliCloud Technology	intellicloudai-001	2019-08-13
31	CSA IntelliCloud Technology	intellicloudai-002	2020-12-17
32	CTBC Bank	ctbcbank-000	2019-06-28
33	Camvi Technologies	camvi-004	2019-07-12
34	Canon Inc	canon-002	2020-12-29
35	Canon Inc	cib-001	2020-08-05
36	China Electronics Import-Export Corp	ceiec-003	2020-01-06
37	China Electronics Import-Export Corp	ceiec-004	2021-01-18
38	China University of Petroleum	upc-001	2019-06-05
39	Chinese Univeristy of Hong Kong	cuhkee-001	2020-03-18
40	Chosun University	chosun-001	2020-07-01
41	Chosun University	chosun-002	2020-11-25
42	Chunghwa Telecom	chtface-002	2019-12-07
43	Chunghwa Telecom	chtface-003	2020-06-24
44	CloudSmart Consulting LLC	csc-001	2020-11-20
45	Cognitec Systems GmbH	cognitec-002	2021-02-24
46	Cortica	cor-001	2020-09-24
47	Cubox	cubox-001	2020-12-07
48	Cybercore	cybercore-000	2020-08-26
49	Cyberlink Corp	cyberlink-005	2020-07-31
50	Cyberlink Corp	cyberlink-006	2021-01-08
51	DSK	dsk-000	2019-06-28
52	Dahua Technology	dahua-005	2020-08-13
53	Dahua Technology	dahua-006	2020-12-30
54	Decatur Industries Inc	decatur-000	2020-08-18
55	Deepglint	deepglint-002	2019-11-15
56	Dermalog	dermalog-007	2020-11-25
57	DiDi ChuXing Technology	didiglobalface-001	2019-10-23

Table 2: List of algorithms included in this report. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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	Developer	Algorithm	Submission Date
58	Expasoft LLC	expasoft-000	2020-01-06
59	Expasoft LLC	expasoft-001	2020-09-03
60	FaceSoft	facesoft-000	2019-07-10
61	Fiberhome Telecommunication Technologies	fiberhome-nanjing-002	2020-08-10
62	Fujitsu Research and Development Center	fujitsulab-001	2020-09-30
63	Fujitsu Research and Development Center	fujitsulab-002	2021-02-24
64	Gemalto Cogent	cogent-005	2020-12-29
65	GeoVision Inc	geo-000	2020-06-29
66	GeoVision Inc	geo-001	2020-10-30
67	Glory	glory-002	2019-11-12
68	Glory	glory-003	2021-01-15
69	Gorilla Technology	gorilla-005	2020-03-11
70	Gorilla Technology	gorilla-006	2020-07-31
71	Guangzhou Pixel Solutions	pixelall-003	2019-10-15
72	Guangzhou Pixel Solutions	pixelall-004	2020-07-02
73	Guangzhou Pixel Solutions	pixelall-005	2021-02-05
74	Hengrui AI Technology	hr-003	2020-09-25
75	Hengrui AI Technology	hr-004	2021-02-10
76	Herta Security	hertasecurity-000	2021-01-05
77	HyperVerge Inc	hv-001	2020-12-13
78	ICM Airport Technics	icm-002	2020-11-13
79	ID3 Technology	id3-005	2020-08-04
80	ID3 Technology	id3-006	2020-12-17
81	ITMO University	itmo-007	2020-01-06
82	IVA Cognitive	ivacognitive-001	2021-01-29
83	Idemia	idemia-006	2020-07-06
84	Idemia	idemia-007	2020-12-04
85	Imageware Systems	iws-000	2020-08-12
86	Imagus Technology Pty	imagus-001	2019-10-22
87	Imperial College London	imperial-002	2019-08-28
88	Incode Technologies Inc	incode-007	2020-08-25
89	Incode Technologies Inc	incode-008	2021-01-19
90	Innef Labs	innefulabs-000	2020-09-04
91	Innovative Technology	innovativetechnologyltd-002	2020-02-26
92	Innovatrics	innovatrics-006	2019-08-13
93	Innovatrics	innovatrics-007	2020-08-19
94	Institute of Computing Technology	icthtc-000	2020-11-29
95	Institute of Information Technologies	iit-002	2019-12-04
96	Institute of Information Technologies	iit-003	2020-12-01
97	Intel Research Group	intelresearch-002	2020-07-24
98	Intel Research Group	intelresearch-003	2021-01-18
99	Intellivision	intellivision-002	2019-08-23
100	IrexAI	irex-000	2020-12-17
101	Kakao Enterprise	kakao-003	2020-02-26
102	Kakao Enterprise	kakao-004	2020-10-28
103	Kedacom International Pte	kedacom-000	2019-06-03
104	Kneron Inc	kneron-005	2020-02-21
105	Kookmin University	kookmin-001	2020-09-28
106	Lomonosov Moscow State University	intsysmsu-002	2020-03-12
107	Lookman Electroplast Industries	lookman-004	2019-06-03
108	Luxand Inc	luxand-000	2019-11-07
109	MVision	mvision-001	2019-11-12
110	Minivision	minivision-000	2020-10-28
111	Mobbeel Solutions	mobbl-000	2021-01-28
112	Momentum Digital	sertis-000	2019-10-07
113	Momentum Digital	sertis-001	2020-07-30
114	Moontime Smart Technology	mt-002	2020-07-02

Table 3: List of algorithms included in this report. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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	Developer	Algorithm	Submission Date
115	Moontime Smart Technology	mt-003	2020-12-22
116	MoreDian Technology	moredian-000	2021-02-24
117	N-Tech Lab	ntech-008	2020-01-06
118	N-Tech Lab	ntechlab-009	2020-12-30
119	NSENSE Corp	nsensecorp-001	2020-10-20
120	Naver Corp	clova-000	2020-10-21
121	Netbridge Technology Incoporation	netbridgetech-001	2020-01-08
122	Netbridge Technology Incoporation	netbridgetech-002	2020-08-11
123	Neurotechnology	neurotechnology-009	2020-07-07
124	Neurotechnology	neurotechnology-010	2020-11-26
125	Nodeflux	nodeflux-002	2019-08-13
126	NotionTag Technologies Private Limited	notiontag-000	2019-06-12
127	Oz Forensics LLC	oz-001	2020-07-29
128	Oz Forensics LLC	oz-002	2021-01-18
129	PXL Vision AG	pxl-001	2020-06-30
130	Panasonic R+D Center Singapore	psl-005	2020-07-06
131	Panasonic R+D Center Singapore	psl-006	2020-11-13
132	Paravision (EverAI)	paravision-004	2019-12-11
133	Paravision (EverAI)	paravision-006	2021-02-01
134	Pensees Pte	pensees-001	2020-08-17
135	Rank One Computing	rankone-009	2020-06-26
136	Rank One Computing	rankone-010	2020-11-05
137	Remark Holdings	remarkai-002	2019-11-21
138	Rokid Corporation	rokid-000	2019-08-01
139	Samsung S1 Corp	s1-001	2019-12-06
140	Satellite Innovation/Eocortex	eocortex-000	2020-08-26
141	Scanovate	scanovate-001	2019-11-12
142	Scanovate	scanovate-002	2020-06-26
143	Securif AI	securifai-001	2020-10-06
144	Sensetime Group	sensetime-003	2019-06-04
145	Sensetime Group	sensetime-004	2020-11-20
146	Shanghai Jiao Tong University	sjtu-002	2020-02-12
147	Shanghai Jiao Tong University	sjtu-003	2020-11-02
148	Shanghai Ulucu Electronics Technology	uluface-002	2019-07-10
149	Shanghai University - Shanghai Film Academy	shu-002	2019-12-10
150	Shanghai University - Shanghai Film Academy	shu-003	2020-06-24
151	Shenzhen AiMall Tech	aimall-002	2020-03-12
152	Shenzhen AiMall Tech	aimall-003	2020-08-12
153	Shenzhen Intellifusion Technologies	intellifusion-002	2020-03-18
154	Shenzhen Univeristy-Macau University of Science and Technology	sztu-000	2020-12-17
155	Staqu Technologies	staqu-000	2020-07-15
156	Star Hybrid Limited	starhybrid-001	2019-06-19
157	Su Zhou NaZhiTianDi intelligent technology	nazhiai-000	2020-06-25
158	Synology Inc	synology-000	2019-10-23
159	Synology Inc	synology-002	2020-08-20
160	TUPU Technology	tuputech-000	2019-10-11
161	Taiwan AI Labs	ailabs-001	2019-12-18
162	Tech5 SA	tech5-004	2020-03-09
163	Tech5 SA	tech5-005	2020-07-24
164	Tencent Deepsea Lab	deepsea-001	2019-06-03
165	Tevian	tevian-005	2019-09-21
166	Tevian	tevian-006	2020-09-11
167	TigerIT Americas LLC	tiger-004	2020-12-01
168	Trueface.ai	trueface-000	2019-10-08
169	Trueface.ai	trueface-001	2020-07-20
170	Universidade de Coimbra	visteam-000	2020-01-14
171	Veridas Digital Authentication Solutions S.L.	veridas-004	2020-07-21

Table 4: List of algorithms included in this report. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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	Developer	Algorithm	Submission Date
172	Veridas Digital Authentication Solutions S.L.	veridas-005	2020-12-15
173	Via Technologies Inc	via-001	2020-01-08
174	Videmo Intelligente Videoanalyse	videmo-000	2019-12-19
175	Videonetics Technology Pvt	videonetics-002	2019-11-21
176	Vietnam Posts and Telecommunications Group	vnpt-001	2021-01-08
177	Viettel Group	vts-000	2020-11-04
178	Vigilant Solutions	vigilantsolutions-008	2020-08-03
179	Vigilant Solutions	vigilantsolutions-009	2020-12-07
180	VinAI Research VietNam	vinai-000	2020-09-24
181	Visage Technologies	visage-000	2020-12-09
182	VisionLabs	visionlabs-009	2020-07-27
183	VisionLabs	visionlabs-010	2021-01-25
184	Vocord	vocord-008	2020-01-31
185	Vocord	vocord-009	2020-12-28
186	Winsense	winsense-001	2019-10-16
187	Winsense	winsense-002	2020-11-20
188	X-Laboratory	x-laboratory-001	2020-01-21
189	Xforward AI Technology	xforwardai-001	2020-09-25
190	Xforward AI Technology	xforwardai-002	2021-02-10
191	Xiamen University	xm-000	2020-10-19
192	YooniK	yoonik-000	2020-06-24
193	YooniK	yoonik-001	2020-10-26
194	Yuan High-Tech Development	yuan-000	2020-06-30
195	Yuan High-Tech Development	yuan-001	2021-01-08
196	iQIYI Inc	iqface-000	2019-06-04
197	iQIYI Inc	iqface-003	2021-02-23
198	iSAP Solution Corporation	isap-002	2020-09-01

Table 5: List of algorithms included in this report. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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5 Results

This section includes accuracy results for the 152 one-to-one verification algorithms listed in Section 4, of which 100 were submitted to FRVT after mid-March 2020 and are labeled in blue in figures and tables throughout this report. We do not include speed and computational resource requirements - they are given in Table 1 in the FRVT 1:1 report. Missing entries generally mean the algorithm was not run on that particular mask variation due to time and resource constraints. The results, which span many pages, are comprised of:

- ▷ **Evolution of algorithm performance with face masks**: This section of figures shows the evolution of performance with face masks for developers that have submitted algorithms since mid-March 2020.
- ▷ **FNMR - summary**: Figure 10 gives a summary of false non-matches rates between pre and post-COVID algorithms with a common type of mask. FNMR values are stated at a fixed threshold calibrated to give FMR = 0.00001 on unmasked images.
- ▷ **FNMR - detailed**: Table 6 tabulates false non-match rates by color, shape, and nose coverage. It includes also FNMR without any mask. FNMR values are stated at a fixed threshold calibrated to give FMR = 0.00001 on unmasked images.
- ▷ **Mask vs. no mask**: The scatter plot in Figure 11 shows variation across all algorithms of FNMR without masks against FNMR with a common type of mask, broken out by pre and post-COVID algorithms.
- ▷ **Mask shape**: The scatter plot in Figure 12 shows for all algorithms the increase in false negative results for wide masks vs. narrower round masks, broken out by pre and post-COVID algorithms.
- ▷ **Mask nose coverage**: The scatter plot in Figure 13 shows for all algorithms the increase in false negative rates for masks that substantially cover the nose and those pulled beneath the nose, broken out by pre and post-COVID algorithms.
- ▷ **FTE**: Table 11 gives empirical failure-to-template results by color, shape, and nose coverage. The table was produced using 10 000 images of each kind of mask.
- ▷ **FTE as contributor to FNMR**: The FNMR results include failure-to-template rates (FTE). Figure 14 shows the proportion of template generation failures, broken out by pre and post-COVID algorithms.
- ▷ **DET - impact of mask nose coverage and shape**: This section of figures shows detection error tradeoff characteristics for each algorithm, across different mask nose coverages and shapes.
- ▷ **DET - impact of mask color**: This section of figures shows detection error tradeoff characteristics for each algorithm, across mask colors.
- ▷ **FNMR and FMR vs. threshold**: This section of figures shows the explicit dependence of false non-match rate and false match rate on threshold.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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The following plots show evolution of accuracy with face masks for developers that have submitted algorithms since mid-March 2020.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 5: Evolution of accuracy with face masks for developers that have submitted algorithms since mid-March 2020. The red line represents results on a masked dataset (masked probe), and for reference, the blue and green lines are results for unmasked datasets. FNMR on masked dataset is for medium, wide, lightblue masks. The horizontal gold line shows where FNMR=10%.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 6: Evolution of accuracy with face masks for developers that have submitted algorithms since mid-March 2020. The red line represents results on a masked dataset (masked probe), and for reference, the blue and green lines are results for unmasked datasets. FNMR on masked dataset is for medium, wide, lightblue masks. The horizontal gold line shows where FNMR=10%.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 7: Evolution of accuracy with face masks for developers that have submitted algorithms since mid-March 2020. The red line represents results on a masked dataset (masked probe), and for reference, the blue and green lines are results for unmasked datasets. FNMR on masked dataset is for medium, wide, lightblue masks. The horizontal gold line shows where FNMR=10%.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 8: Evolution of accuracy with face masks for developers that have submitted algorithms since mid-March 2020. The red line represents results on a masked dataset (masked probe), and for reference, the blue and green lines are results for unmasked datasets. FNMR on masked dataset is for medium, wide, lightblue masks. The horizontal gold line shows where FNMR=10%.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 9: Evolution of accuracy with face masks for developers that have submitted algorithms since mid-March 2020. The red line represents results on a masked dataset (masked probe), and for reference, the blue and green lines are results for unmasked datasets. FNMR on masked dataset is for medium, wide, lightblue masks. The horizontal gold line shows where FNMR=10%.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 10: For each algorithm, each dot shows FNMR @ FMR=0.00001, where the threshold is set for FMR on unmasked probe images. The results are for when the probe is masked, and the enrollment image is unmasked. Gray dots represent results for algorithms submitted prior to mid-March 2020 (pre-COVID), and blue dots represent algorithms submitted thereafter (post-COVID).

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

	Algorithm	NOT MASKED	MASK COLOR = LIGHTBLUE						COLOR = BLACK			COLOR = RED	COLOR = WHITE
	Name		SHAPE = WIDE			SHAPE = ROUND			SHAPE = WIDE			SHAPE = WIDE	SHAPE = WIDE
	COVERAGE		LO	MED	HI	LO	MED	HI	LO	MED	HI	MED	MED
1	3divi-004	0.0130^{117}	0.4123^{109}	0.6760^{156}	-	-	-	-	-	-	-	-	-
2	3divi-005	0.0088^{88}	0.1006^{83}	0.2895^{100}	0.3670^{59}	0.0783^{47}	0.1918^{49}	-	0.1185^{53}	0.3057^{60}	0.4272^{49}	-	-
3	acer-000	0.8432^{190}	0.9995^{133}	0.9999^{192}	-	-	-	-	-	-	-	-	-
4	acer-001	0.0219^{141}	0.2587^{103}	0.5835^{144}	0.6719^{71}	0.1536^{52}	0.4325^{57}	0.6831^{45}	0.3132^{59}	0.6304^{69}	0.7143^{54}	-	-
5	acisw-003	0.7160^{187}	0.9810^{126}	0.9970^{182}	0.9970^{80}	-	-	-	0.9883^{67}	0.9997^{80}	-	-	-
6	acisw-006	0.7227^{188}	-	0.9993^{186}	-	-	-	-	-	-	-	-	-
7	adera-002	0.0073^{72}	-	0.1461^{61}	-	-	-	-	-	-	-	-	-
8	advance-002	0.0328^{151}	-	0.2351^{85}	-	-	-	-	-	-	-	0.2333^{58}	-
9	aifirst-001	0.0079^{77}	0.0778^{70}	0.2567^{89}	-	-	-	-	-	-	-	0.2624^{61}	-
10	aigen-001	0.0159^{133}	0.1268^{91}	0.3790^{117}	0.4432^{66}	-	0.2880^{56}	-	0.1956^{57}	0.4761^{65}	0.6329^{53}	0.3189^{63}	0.3754^{47}
11	ailabs-001	0.0243^{142}	-	0.6792^{157}	-	-	-	-	-	-	-	-	-
12	aimall-002	0.0133^{121}	-	0.3919^{118}	-	-	-	-	-	-	-	-	-
13	aimall-003	0.0045^{30}	0.0188^{26}	0.0781^{38}	0.1325^{34}	0.0175^{16}	0.0524^{27}	0.1021^{23}	0.0223^{20}	0.0913^{29}	0.1577^{27}	0.0738^{24}	0.0800^{28}
14	aiunionface-000	0.0094^{92}	0.0917^{79}	0.2935^{101}	-	-	-	-	-	-	-	-	-
15	alleyes-000	0.0044^{28}	-	0.1038^{51}	-	0.0181^{20}	0.0542^{29}	0.1050^{24}	0.0262^{27}	0.1287^{37}	0.1991^{30}	0.1066^{35}	0.1098^{34}
16	alphaface-002	1.0000^{197}	1.0000^{135}	1.0000^{195}	-	-	-	-	-	-	-	-	-
17	androvideo-000	0.0333^{152}	-	0.6488^{151}	-	-	-	-	-	-	-	-	-
18	anke-005	0.0062^{60}	0.0671^{65}	0.3207^{108}	-	-	-	-	-	-	-	-	-
19	antheus-000	0.7319^{189}	0.9994^{132}	0.9999^{191}	-	-	-	-	-	-	-	-	-
20	antheus-001	0.6608^{184}	0.9988^{131}	0.9998^{190}	0.9998^{81}	0.9993^{60}	0.9998^{69}	0.9998^{51}	0.9993^{69}	0.9998^{81}	0.9998^{64}	-	-
21	anyvision-005	0.0037^{12}	0.0119^{11}	0.0548^{22}	-	-	-	-	-	-	-	-	-
22	asusaics-000	0.0090^{90}	-	0.3616^{116}	-	-	-	-	-	-	-	-	-
23	aware-005	0.0308^{150}	0.4962^{113}	0.8876^{166}	-	-	-	-	-	-	-	-	-
24	awiros-001	0.1233^{172}	0.6823^{120}	0.8635^{164}	-	-	-	-	-	-	-	-	-
25	awiros-002	0.2283^{177}	0.6356^{117}	0.8671^{165}	0.9221^{76}	-	0.7932^{64}	-	0.7703^{65}	0.9068^{77}	0.9379^{58}	0.8628^{68}	0.8508^{50}
26	ayftech-001	0.0828^{168}	0.6740^{118}	0.9132^{170}	0.9333^{78}	0.5865^{58}	0.8519^{66}	0.9538^{50}	0.7540^{64}	0.9048^{76}	0.9705^{62}	-	-
27	bioidtechswiss-001	0.0037^{10}	0.0225^{32}	0.0945^{44}	0.1519^{38}	0.0180^{19}	0.0524^{28}	0.1070^{26}	0.0254^{25}	0.0986^{32}	0.1571^{25}	0.0979^{32}	0.0945^{33}
28	bioidtechswiss-002	0.0047^{35}	-	0.1021^{49}	-	-	-	-	-	-	-	-	-
29	bresee-000	0.9802^{191}	0.9968^{130}	0.9989^{184}	-	-	-	-	0.9975^{68}	-	-	-	-
30	bresee-001	0.0067^{67}	-	0.1382^{57}	-	-	-	-	-	-	-	-	-
31	camvi-004	0.0063^{61}	0.0697^{67}	0.2179^{82}	-	-	-	-	-	-	-	0.3337^{64}	-
32	canon-002	0.0033^{4}	0.0125^{13}	0.0565^{23}	0.0888^{19}	-	-	-	0.0138^{11}	0.0629^{19}	-	0.0729^{23}	0.0602^{18}
33	ceiec-003	0.0134^{123}	-	0.1230^{52}	-	-	-	-	-	-	-	-	-
34	ceiec-004	0.0051^{44}	-	0.0618^{28}	-	-	-	-	-	-	-	-	-
35	chosun-001	0.0582^{166}	0.5759^{114}	0.9091^{168}	0.9263^{77}	0.5021^{57}	0.8531^{67}	0.9278^{48}	0.5158^{61}	0.9031^{75}	0.9409^{59}	-	-
36	chosun-002	0.0266^{145}	-	0.5807^{143}	-	-	0.4707^{58}	-	-	-	-	-	-
37	chtface-002	0.0108^{105}	0.1423^{93}	0.4303^{126}	-	-	-	-	-	-	-	-	-
38	chtface-003	0.0084^{83}	0.1008^{84}	0.3201^{107}	-	0.0774^{46}	0.2393^{54}	0.4387^{41}	0.1232^{55}	0.3309^{62}	0.5044^{51}	-	-
39	cib-001	0.0036^{9}	0.0149^{19}	0.0731^{36}	0.1095^{29}	0.0125^{7}	0.0336^{13}	0.0690^{8}	0.0175^{15}	0.0731^{24}	0.1137^{12}	0.0863^{28}	0.0779^{27}
40	clova-000	0.0087^{86}	0.0881^{77}	0.3411^{110}	0.5017^{67}	-	0.1752^{47}	-	0.0950^{51}	0.3051^{59}	0.4211^{48}	-	-
41	cogent-005	0.0072^{70}	0.0260^{37}	0.0351^{11}	0.0631^{13}	-	-	-	0.0252^{24}	0.0349^{11}	-	0.0335^{11}	0.0363^{10}
42	cognitec-002	0.0130^{116}	-	0.2201^{83}	-	-	-	-	-	-	-	-	-
43	cor-001	0.0053^{47}	0.0504^{58}	0.1802^{74}	0.2470^{50}	0.0300^{29}	0.0864^{32}	-	0.0364^{34}	0.1328^{38}	0.1828^{28}	0.1527^{42}	-
44	csc-001	0.9993^{194}	-	1.0000^{196}	-	-	-	-	-	-	-	-	-
45	ctbcbank-000	0.0133^{119}	0.1594^{100}	0.7448^{161}	-	-	-	-	-	-	-	-	-

Table 6: This table summarizes False Non-Match Rate (FNMR) on unmasked and masked probe images. FNMR is the proportion of mated comparisons below a threshold set to achieve FMR=1e-05 on unmasked probe images. False Match Rate (FMR) is the proportion of impostor comparisons at or above that threshold. The red superscripts give rank over all algorithms in that column. Missing entries generally mean the algorithm was not run on that particular mask variation due to time and resource constraints. Algorithms with FTE=1.00 were not run at all. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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	Algorithm	NOT MASKED	MASK COLOR = LIGHTBLUE						COLOR = BLACK			COLOR = RED	COLOR = WHITE
	Name		SHAPE = WIDE			SHAPE = ROUND			SHAPE = WIDE			SHAPE = WIDE	SHAPE = WIDE
	COVERAGE		LO	MED	HI	LO	MED	HI	LO	MED	HI	MED	MED
46	cubox-001	0.0042[22]	-	0.1037[50]	-	-	-	-	-	-	-	-	-
47	cuhkee-001	0.0041[21]	0.0143[18]	0.0572[24]	0.0963[26]	0.0143[11]	0.0333[12]	0.0715[9]	0.0164[13]	0.0652[20]	0.1193[13]	-	-
48	cybercore-000	0.1096[170]	-	0.9097[169]	-	-	-	-	-	-	-	-	-
49	cyberlink-005	0.0055[48]	0.0376[46]	0.1374[56]	0.1808[41]	0.0388[40]	0.1018[36]	0.2695[37]	0.0451[41]	0.1610[44]	0.3991[46]	0.4285[66]	0.1316[35]
50	cyberlink-006	0.0040[19]	0.0190[27]	0.0684[34]	0.0944[25]	-	-	-	-	-	-	0.1347[39]	0.0723[24]
51	dahua-005	0.0034[6]	0.0155[20]	0.0660[32]	0.1135[32]	0.0127[9]	0.0340[14]	0.0829[14]	0.0137[10]	0.0601[18]	0.1019[10]	0.0579[16]	0.0613[19]
52	dahua-006	0.0031[3]	0.0100[9]	0.0399[13]	0.0701[15]	-	-	-	0.0098[6]	0.0370[12]	-	0.0350[12]	0.0393[11]
53	decatur-000	0.0384[156]	-	0.6440[149]	-	-	-	-	-	-	-	-	-
54	deepglint-002	0.0039[18]	0.0077[4]	0.0237[3]	0.0455[5]	0.0078[2]	0.0141[3]	0.0292[2]	0.0083[3]	0.0254[2]	0.0513[4]	0.0234[2]	0.0248[2]
55	deepsea-001	0.0110[108]	0.1218[88]	0.3094[105]	0.3778[62]	0.0922[50]	0.2217[52]	0.4469[44]	-	-	-	-	-
56	dermalog-007	0.1298[173]	-	0.9617[173]	-	-	-	-	-	-	-	-	-
57	didiglobalface-001	0.0050[40]	-	0.0986[45]	0.1517[36]	0.0255[23]	0.0515[25]	0.0979[21]	0.0291[29]	0.1033[34]	0.1558[24]	0.1241[36]	0.1655[41]
58	dsk-000	0.1961[174]	0.9108[123]	0.9929[179]	-	-	-	-	-	-	-	-	-
59	eocortex-000	0.1187[171]	-	0.9694[175]	-	-	-	-	-	-	-	-	-
60	ercacat-001	0.0096[94]	0.0187[25]	0.0616[27]	0.0994[27]	0.0173[15]	0.0357[15]	0.0728[11]	0.0200[18]	0.0663[22]	0.1114[11]	0.0679[21]	0.0648[21]
61	expasoft-000	0.0519[164]	0.3186[106]	0.6796[158]	-	-	-	-	-	-	-	-	-
62	expasoft-001	0.0492[162]	-	0.6414[148]	-	-	-	-	-	-	-	-	-
63	facesoft-000	0.0057[53]	0.0397[49]	0.1428[59]	-	-	-	-	-	0.1573[43]	-	0.1446[41]	0.1428[38]
64	fiberhome-nanjing-002	0.0133[120]	0.1554[99]	0.4734[133]	-	-	-	-	0.1641[56]	-	-	-	-
65	fujitsulab-001	0.0078[76]	0.0330[41]	0.0427[15]	0.0608[11]	0.0311[30]	0.0363[16]	0.0666[7]	0.0407[37]	0.0985[30]	0.1525[22]	0.0898[30]	0.0776[26]
66	fujitsulab-002	0.0106[104]	0.0433[50]	0.0677[33]	-	-	-	-	-	-	-	-	-
67	geo-000	0.0902[169]	0.5765[115]	0.9236[171]	-	-	0.8082[65]	0.9311[49]	0.6214[63]	0.9428[78]	0.9541[60]	-	-
68	geo-001	0.0051[41]	0.0229[34]	0.1008[48]	-	-	-	-	-	-	-	0.0975[31]	-
69	glory-002	0.0109[107]	-	0.2729[97]	-	-	-	-	-	-	-	-	-
70	glory-003	0.0081[79]	-	0.2370[87]	-	-	-	-	-	-	-	-	-
71	gorilla-005	0.0117[111]	0.1463[95]	0.5037[137]	-	-	-	-	-	-	-	-	-
72	gorilla-006	0.0085[84]	0.0495[57]	0.1805[75]	0.2883[54]	0.0385[39]	0.1134[43]	-	-	0.2260[53]	0.3419[42]	0.1980[53]	-
73	hertasecurity-000	0.0464[161]	-	0.5644[141]	-	-	-	-	-	-	-	-	-
74	hr-003	0.0049[39]	0.0133[15]	0.0419[14]	0.0613[12]	0.0122[5]	0.0247[9]	0.0475[6]	0.0142[12]	0.0527[15]	0.0914[8]	0.0542[15]	0.0476[14]
75	hr-004	0.0048[38]	0.0086[6]	0.0240[4]	0.0373[2]	-	0.0135[2]	-	-	0.0313[8]	-	0.0302[6]	0.0282[6]
76	hv-001	1.0000[198]	-	1.0000[198]	-	-	-	-	-	-	-	-	-
77	icm-002	0.0150[130]	-	0.3484[113]	-	-	-	-	-	-	-	-	-
78	icthtc-000	0.0197[138]	-	0.4887[134]	-	-	-	-	-	-	-	-	-
79	id3-005	0.0145[128]	0.1181[87]	0.3540[115]	-	-	-	-	-	-	-	-	-
80	id3-006	0.0066[65]	-	0.1800[73]	-	-	-	-	-	-	-	-	-
81	idemia-006	0.0048[37]	0.0327[40]	0.0539[19]	0.0911[21]	0.0322[31]	0.0440[22]	0.0946[18]	0.0359[32]	0.0714[23]	0.1489[20]	0.0607[19]	0.0564[17]
82	idemia-007	0.0035[8]	-	0.0988[46]	-	-	-	-	-	-	-	-	-
83	iit-002	0.0141[126]	-	0.3078[104]	-	-	-	-	-	-	-	-	-
84	iit-003	0.0080[78]	-	0.1975[81]	-	-	-	-	-	-	-	-	-
85	imagus-001	0.0276[146]	0.3488[108]	0.6510[152]	-	-	-	-	-	-	-	-	-
86	imperial-002	0.0055[49]	0.0320[39]	0.1350[55]	0.1972[42]	0.0258[24]	0.0775[31]	0.1556[27]	0.0359[33]	0.1510[42]	0.2302[35]	0.1533[43]	0.1432[39]
87	incode-007	0.0058[54]	0.0470[54]	0.1790[72]	0.2393[49]	0.0323[32]	0.1040[38]	0.1779[29]	0.0561[45]	0.1891[50]	0.2754[38]	0.1978[52]	-
88	incode-008	0.0039[16]	-	0.1004[47]	-	-	-	-	-	-	-	-	-
89	innefulabs-000	0.0155[132]	-	0.2971[102]	-	-	-	-	-	-	-	-	-
90	innovativetechnologyltd-002	0.0251[143]	0.2701[104]	0.6454[150]	-	-	-	-	-	-	-	-	-

Table 7: This table summarizes False Non-Match Rate (FNMR) on unmasked and masked probe images. FNMR is the proportion of mated comparisons below a threshold set to achieve FMR=1e-05 on unmasked probe images. False Match Rate (FMR) is the proportion of impostor comparisons at or above that threshold. The red superscripts give rank over all algorithms in that column. Missing entries generally mean the algorithm was not run on that particular mask variation due to time and resource constraints. Algorithms with FTE=1.00 were not run at all. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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	Algorithm	NOT MASKED	MASK COLOR = LIGHTBLUE						COLOR = BLACK			COLOR = RED	COLOR = WHITE
	Name		SHAPE = WIDE			SHAPE = ROUND			SHAPE = WIDE			SHAPE = WIDE	SHAPE = WIDE
	COVERAGE		LO	MED	HI	LO	MED	HI	LO	MED	HI	MED	MED
91	innovatrics-006	0.0059[57]	0.0543[64]	0.2210[84]	0.3118[55]	0.0369[35]	0.1109[41]	0.1984[31]	0.0557[44]	0.1909[51]	0.2764[39]	0.1936[51]	-
92	innovatrics-007	0.0051[42]	-	0.1537[64]	-	-	-	-	-	-	-	-	-
93	intellicloudai-001	0.0095[93]	0.1044[85]	0.4394[128]	-	-	-	-	-	-	-	-	-
94	intellicloudai-002	0.0082[80]	0.0226[33]	0.0470[17]	0.0792[17]	-	-	-	-	-	-	0.0535[14]	-
95	intellifusion-002	0.0056[52]	0.0539[62]	0.1690[69]	-	-	-	-	-	0.1822[49]	-	0.2556[60]	0.2119[45]
96	intellivision-002	0.0463[160]	0.5999[116]	0.9028[167]	-	-	-	-	-	-	-	-	-
97	intelresearch-002	0.0111[109]	0.0458[52]	0.1598[66]	0.2085[44]	0.0384[38]	0.1025[37]	0.2568[36]	0.0598[46]	0.1764[48]	0.2583[37]	0.1857[49]	0.1595[40]
98	intelresearch-003	0.0071[68]	-	0.0930[42]	-	-	-	-	-	-	-	-	-
99	intsysmsu-002	0.0089[89]	0.0827[73]	0.3138[106]	-	-	-	-	-	-	-	-	-
100	iqface-000	0.0128[115]	0.0885[78]	0.2867[99]	-	-	-	-	-	-	-	-	-
101	iqface-003	0.0099[99]	0.0254[36]	0.0592[26]	-	-	-	-	-	-	-	-	-
102	irex-000	0.0046[34]	-	0.1491[63]	-	-	-	-	-	-	-	-	-
103	isap-002	0.0094[91]	-	0.4090[122]	-	-	-	-	-	-	-	-	-
104	itmo-007	0.0098[98]	0.0840[74]	0.2685[94]	-	-	-	-	-	-	-	-	-
105	ivacognitive-001	0.0096[95]	-	0.4245[125]	-	-	-	-	-	-	-	-	-
106	iws-000	0.6797[185]	0.9960[128]	0.9997[189]	-	-	-	-	-	-	-	-	-
107	kakao-003	0.0170[135]	0.1541[98]	0.4123[123]	-	-	-	-	-	-	-	-	-
108	kakao-004	0.0125[114]	-	0.1741[70]	-	-	-	-	-	-	-	-	-
109	kedacom-000	0.0391[157]	0.3444[107]	0.6188[147]	0.6848[72]	0.2663[54]	0.5975[61]	-	-	-	-	-	-
110	kneron-005	0.0296[149]	-	0.4567[130]	-	-	-	-	-	-	-	-	-
111	kookmin-001	0.0357[154]	0.4479[111]	0.7378[160]	0.8497[74]	0.2933[55]	0.6006[62]	-	0.4160[60]	0.7365[73]	0.8917[56]	0.8642[69]	0.7432[49]
112	lookman-004	0.0398[158]	-	0.6520[153]	-	-	-	-	-	-	-	-	-
113	luxand-000	0.2167[176]	0.9732[125]	0.9988[183]	-	-	-	-	-	-	-	-	-
114	minivision-000	0.0043[24]	0.0165[21]	0.0763[37]	-	-	-	-	-	-	-	-	-
115	mobbl-000	0.4931[181]	-	0.9965[181]	-	-	-	-	-	-	-	-	-
116	moredian-000	0.9949[192]	-	0.9996[188]	-	-	-	-	-	-	-	-	-
117	mt-002	0.0043[25]	0.0215[29]	0.0713[35]	0.1114[30]	0.0162[13]	0.0404[19]	0.0777[13]	0.0225[21]	0.0877[28]	0.1399[19]	0.0804[26]	0.0757[25]
118	mt-003	0.0034[7]	0.0076[2]	0.0237[2]	0.0447[4]	0.0078[1]	0.0156[5]	0.0324[4]	0.0082[2]	0.0254[3]	0.0482[2]	0.0235[3]	0.0268[3]
119	mvision-001	0.0137[124]	-	0.3987[119]	-	-	-	-	-	-	-	-	-
120	nazhiai-000	0.0043[26]	0.0184[23]	0.0835[39]	0.1318[33]	0.0156[12]	0.0463[23]	0.0947[19]	0.0177[16]	0.0764[25]	0.1271[16]	0.0792[25]	0.0854[31]
121	netbridgetech-001	0.2673[179]	0.8940[122]	0.9878[178]	-	-	-	-	-	-	-	-	-
122	netbridgetech-002	0.0083[82]	0.0781[71]	0.2723[96]	0.3522[57]	0.0528[42]	0.1551[45]	-	0.0875[50]	0.2863[58]	0.4151[47]	-	-
123	neurotechnology-009	0.0051[43]	0.0351[44]	0.1768[71]	0.2537[52]	0.0268[25]	0.0953[34]	0.2034[32]	0.0414[38]	0.2180[52]	0.3192[41]	0.1658[45]	-
124	neurotechnology-010	0.0048[36]	0.0384[47]	0.0591[25]	0.1070[28]	-	-	-	-	-	-	0.0598[17]	0.0616[20]
125	nodeflux-002	0.0424[159]	0.4177[110]	0.7307[159]	-	-	-	-	-	-	-	-	-
126	notiontag-000	0.6814[186]	0.9966[129]	0.9992[185]	-	-	-	-	-	-	-	-	-
127	nsensecorp-001	1.0000[196]	1.0000[136]	1.0000[197]	1.0000[82]	-	1.0000[70]	-	1.0000[70]	1.0000[82]	1.0000[65]	1.0000[71]	1.0000[52]
128	ntechlab-008	0.0033[5]	0.0179[22]	0.0642[30]	0.1126[31]	0.0137[10]	0.0413[21]	0.0953[20]	0.0208[19]	0.0842[27]	0.1348[18]	0.0645[20]	0.0840[30]
129	ntechlab-009	0.0260[144]	0.0361[45]	0.0652[31]	0.0851[18]	-	-	-	-	-	-	0.0878[29]	0.0550[16]
130	oz-001	0.0152[131]	0.1969[102]	0.5887[145]	-	-	-	-	-	-	-	-	-
131	oz-002	0.0105[102]	-	0.1937[80]	-	-	-	-	-	-	-	-	-
132	paravision-004	0.0088[87]	0.0124[12]	0.0281[7]	0.0476[7]	0.0125[8]	0.0181[7]	0.0313[3]	0.0135[8]	0.0327[10]	0.0581[6]	0.0332[10]	0.0313[8]
133	paravision-006	0.0071[69]	0.0109[10]	0.0289[8]	0.0515[9]	-	0.0192[8]	-	-	0.0297[5]	-	0.0310[8]	0.0329[9]
134	pensees-001	0.0106[103]	0.0309[38]	0.0461[16]	0.0921[24]	0.0326[34]	0.0413[20]	0.0893[16]	0.0333[31]	0.0579[17]	0.1217[14]	0.1338[38]	0.0462[13]
135	pixelall-003	0.0086[85]	0.0746[68]	0.2680[92]	-	-	-	-	-	-	-	-	-

Table 8: This table summarizes False Non-Match Rate (FNMR) on unmasked and masked probe images. FNMR is the proportion of mated comparisons below a threshold set to achieve FMR=1e-05 on unmasked probe images. False Match Rate (FMR) is the proportion of impostor comparisons at or above that threshold. The red superscripts give rank over all algorithms in that column. Missing entries generally mean the algorithm was not run on that particular mask variation due to time and resource constraints. Algorithms with FTE=1.00 were not run at all. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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	Algorithm	NOT MASKED	MASK COLOR = LIGHTBLUE						COLOR = BLACK			COLOR = RED	COLOR = WHITE
	Name		SHAPE = WIDE			SHAPE = ROUND			SHAPE = WIDE			SHAPE = WIDE	SHAPE = WIDE
	COVERAGE		LO	MED	HI	LO	MED	HI	LO	MED	HI	MED	MED
136	pixelall-004	0.0078^{75}	0.0524^{61}	0.1676^{68}	0.2141^{46}	0.0495^{41}	0.1131^{42}	0.2254^{35}	0.0849^{48}	0.2461^{54}	0.3158^{40}	0.3148^{62}	0.1791^{43}
137	pixelall-005	0.0065^{63}	-	0.1383^{58}	-	-	-	-	-	-	-	-	-
138	psl-005	0.0056^{51}	0.0433^{51}	0.1904^{78}	0.2576^{53}	0.0292^{28}	0.1054^{40}	0.2210^{34}	0.0433^{40}	0.1681^{46}	0.2223^{33}	0.1851^{48}	-
139	psl-006	0.0148^{129}	0.0245^{35}	0.0631^{29}	0.0914^{23}	-	-	-	-	0.0568^{16}	-	0.0692^{22}	0.0692^{23}
140	ptakuratsatu-000	0.0059^{58}	0.0542^{63}	0.1897^{77}	0.2485^{51}	0.0380^{37}	0.1047^{39}	-	0.0502^{42}	0.1652^{45}	0.2360^{36}	0.2003^{54}	-
141	pxl-001	0.0511^{163}	0.4879^{112}	0.8183^{162}	0.8754^{75}	0.3675^{56}	0.7276^{63}	0.9182^{47}	0.5416^{62}	0.8047^{74}	0.8718^{55}	-	-
142	rankone-009	0.0065^{62}	0.0951^{80}	0.3450^{111}	0.3709^{60}	0.1004^{51}	0.2244^{53}	0.4467^{43}	-	0.5624^{66}	0.9343^{57}	-	-
143	rankone-010	0.0065^{64}	0.1514^{97}	0.5452^{139}	0.3915^{63}	-	0.4726^{59}	-	0.2298^{58}	0.7196^{72}	0.9647^{61}	0.5811^{67}	0.5719^{48}
144	remarkai-002	0.0073^{71}	0.0685^{66}	0.2352^{86}	-	-	-	-	-	-	-	0.2406^{59}	-
145	rokid-000	0.0117^{110}	0.1448^{94}	0.4346^{127}	-	-	-	-	-	-	-	-	-
146	s1-001	0.0277^{147}	0.6776^{119}	0.9459^{172}	-	-	-	-	-	-	-	-	-
147	scanovate-001	0.2403^{178}	-	0.5973^{146}	-	-	-	-	-	-	-	-	-
148	scanovate-002	0.0119^{112}	0.1304^{92}	0.4006^{121}	0.5142^{68}	0.0757^{45}	0.2206^{51}	0.3622^{39}	0.1215^{54}	0.3172^{61}	0.4633^{50}	-	-
149	securifai-001	0.5185^{182}	0.9381^{124}	0.9835^{177}	0.9880^{79}	0.9146^{59}	0.9788^{68}	-	0.9352^{66}	0.9853^{79}	0.9949^{63}	0.9867^{70}	0.9829^{51}
150	sensetime-003	0.0045^{33}	0.0185^{24}	0.0544^{20}	0.0912^{22}	0.0221^{22}	0.0365^{17}	0.0739^{12}	0.0232^{23}	0.0654^{21}	0.1230^{15}	-	-
151	sensetime-004	0.0039^{17}	0.0129^{14}	0.0277^{6}	0.0469^{6}	-	-	-	0.0165^{14}	0.0301^{6}	-	0.0309^{7}	-
152	sertis-000	0.0066^{66}	0.0751^{69}	0.2685^{93}	-	-	-	-	-	-	-	-	-
153	sertis-001	0.0179^{136}	-	0.5774^{142}	0.6655^{70}	-	-	-	-	0.5632^{67}	-	-	-
154	shu-002	1.0000^{195}	-	1.0000^{194}	-	-	-	-	-	-	-	-	-
155	shu-003	0.0053^{46}	0.0465^{53}	0.1839^{76}	0.2148^{47}	0.0379^{36}	0.1148^{44}	0.2196^{33}	0.0422^{39}	0.1702^{47}	0.2210^{32}	0.1901^{50}	-
156	sjtu-002	0.0052^{45}	0.0475^{56}	0.1912^{79}	-	-	-	-	-	-	-	0.2090^{55}	-
157	sjtu-003	0.0043^{23}	0.0340^{43}	0.1239^{53}	0.1609^{39}	-	-	-	0.0276^{28}	0.1008^{33}	0.1539^{23}	0.1333^{37}	0.1323^{36}
158	staqu-000	0.0108^{106}	0.1251^{90}	0.3537^{114}	0.4429^{65}	0.0913^{49}	0.2434^{55}	0.4447^{42}	-	0.3862^{64}	0.6319^{52}	-	-
159	starhybrid-001	0.0104^{101}	0.1923^{101}	0.5033^{136}	-	-	-	-	-	-	-	-	-
160	synology-000	0.0123^{113}	-	0.4459^{129}	-	-	-	-	-	-	-	-	-
161	synology-002	0.0100^{100}	-	0.4666^{131}	-	-	-	-	-	-	-	-	-
162	sztu-000	0.0074^{74}	-	0.1432^{60}	-	-	-	-	-	-	-	-	-
163	tech5-004	0.0045^{31}	0.0218^{30}	0.0839^{40}	0.1389^{35}	0.0172^{14}	0.0464^{24}	0.0905^{17}	0.0228^{22}	0.0818^{26}	0.1288^{17}	0.0826^{27}	0.0830^{29}
164	tech5-005	0.0037^{11}	0.0224^{31}	0.0941^{43}	0.1518^{37}	0.0180^{18}	0.0524^{26}	0.1066^{25}	0.0254^{26}	0.0986^{31}	0.1571^{26}	0.0979^{33}	0.0945^{32}
165	tevian-005	0.0061^{59}	0.0961^{81}	0.5044^{138}	-	-	-	-	-	0.6178^{68}	-	-	-
166	tevian-006	0.0038^{15}	0.0140^{16}	0.0547^{21}	0.0894^{20}	0.0184^{21}	0.0325^{11}	0.1013^{22}	0.0183^{17}	0.1281^{36}	0.1507^{21}	0.1437^{40}	0.0659^{22}
167	tiger-004	0.0357^{155}	-	0.6592^{154}	-	-	-	-	-	-	-	-	-
168	trueface-000	0.0143^{127}	0.1512^{96}	0.4164^{124}	-	-	-	-	-	-	-	-	-
169	trueface-001	0.0097^{97}	0.0978^{82}	0.3483^{112}	0.4310^{64}	0.0710^{44}	0.2064^{50}	0.3846^{40}	-	0.3752^{63}	-	-	-
170	tuputech-000	0.2014^{175}	0.8743^{121}	0.9731^{176}	-	-	-	-	-	-	-	-	-
171	uluface-002	0.0073^{73}	0.0796^{72}	0.2450^{88}	-	-	-	-	-	-	-	0.3939^{65}	-
172	upc-001	0.0167^{134}	-	0.4723^{132}	-	-	-	-	-	-	-	-	-
173	veridas-004	0.0202^{140}	0.2783^{105}	0.6697^{155}	0.7334^{73}	0.1941^{53}	0.5025^{60}	0.6906^{46}	-	0.6592^{71}	-	-	-
174	veridas-005	0.0354^{153}	-	0.8325^{163}	-	-	-	-	-	-	-	-	-
175	via-001	0.0097^{96}	0.1234^{89}	0.3406^{109}	-	-	-	-	-	-	-	-	-
176	videmo-000	0.0140^{125}	-	0.5509^{140}	-	-	-	-	-	-	-	-	-
177	videonetics-002	0.6032^{183}	0.9941^{127}	0.9996^{187}	-	-	-	-	-	-	-	-	-
178	vigilantsolutions-008	0.0195^{137}	-	0.4970^{135}	0.5521^{69}	-	-	-	-	0.6365^{70}	-	-	-
179	vigilantsolutions-009	0.0133^{122}	-	0.3998^{120}	-	-	-	-	-	-	-	-	-
180	vinai-000	0.0056^{50}	0.0388^{48}	0.1587^{65}	0.2130^{45}	0.0286^{27}	0.0965^{35}	-	0.0364^{35}	0.1487^{40}	0.2024^{31}	0.1691^{46}	0.1710^{42}

Table 9: This table summarizes False Non-Match Rate (FNMR) on unmasked and masked probe images. FNMR is the proportion of mated comparisons below a threshold set to achieve FMR=1e-05 on unmasked probe images. False Match Rate (FMR) is the proportion of impostor comparisons at or above that threshold. The red superscripts give rank over all algorithms in that column. Missing entries generally mean the algorithm was not run on that particular mask variation due to time and resource constraints. Algorithms with FTE=1.00 were not run at all. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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	Algorithm	NOT MASKED	MASK COLOR = LIGHTBLUE						COLOR = BLACK			COLOR = RED	COLOR = WHITE
	Name		SHAPE = WIDE			SHAPE = ROUND			SHAPE = WIDE			SHAPE = WIDE	SHAPE = WIDE
	COVERAGE		LO	MED	HI	LO	MED	HI	LO	MED	HI	MED	MED
181	visage-000	0.0674[167]	-	0.9626[174]	-	-	-	-	-	-	-	-	-
182	visionlabs-009	0.0028[2]	0.0095[8]	0.0355[12]	0.0667[14]	0.0123[6]	0.0307[10]	0.0719[10]	0.0123[7]	0.0423[13]	0.0943[9]	0.0399[13]	0.0395[12]
183	visionlabs-010	0.0027[1]	0.0076[3]	0.0342[10]	0.0513[8]	-	-	-	-	0.0313[9]	-	0.0244[4]	0.0269[4]
184	visteam-000	0.9960[193]	1.0000[134]	1.0000[193]	-	-	-	-	-	-	-	-	-
185	vnpt-001	0.3344[180]	-	0.9942[180]	-	-	-	-	-	-	-	-	-
186	vocord-008	0.0038[14]	0.0140[17]	0.0500[18]	0.0762[16]	0.0176[17]	0.0393[18]	0.0892[15]	0.0135[9]	0.0459[14]	0.0771[7]	0.0607[18]	0.0482[15]
187	vocord-009	0.0045[32]	0.0086[5]	0.0261[5]	0.0438[3]	0.0086[3]	0.0151[4]	0.0271[1]	0.0093[5]	0.0289[4]	0.0506[3]	0.0275[5]	0.0271[5]
188	vts-000	0.0199[139]	0.0870[75]	0.2755[98]	0.3566[58]	-	-	-	0.0858[49]	0.2584[57]	0.3898[45]	0.2249[56]	0.2976[46]
189	winsense-001	0.0058[55]	0.0473[55]	0.1626[67]	0.2244[48]	0.0325[33]	0.0946[33]	0.1853[30]	0.0406[36]	0.1471[39]	0.2231[34]	0.1622[44]	0.1843[44]
190	winsense-002	0.0044[29]	0.0213[28]	0.0846[41]	-	-	-	-	-	-	-	0.1014[34]	-
191	x-laboratory-001	0.0058[56]	0.0517[60]	0.2569[90]	-	-	-	-	-	-	-	0.2333[57]	-
192	xforwardai-001	0.0041[20]	0.0087[7]	0.0289[9]	0.0536[10]	0.0087[4]	0.0180[6]	0.0377[5]	0.0090[4]	0.0303[7]	0.0544[5]	0.0313[9]	0.0294[7]
193	xforwardai-002	0.0037[13]	0.0062[1]	0.0159[1]	0.0338[1]	-	0.0119[1]	-	0.0062[1]	0.0153[1]	0.0310[1]	0.0156[1]	0.0162[1]
194	xm-000	0.0044[27]	0.0334[42]	0.1255[54]	0.1682[40]	0.0275[26]	0.0774[30]	0.1648[28]	0.0324[30]	0.1274[35]	0.1839[29]	0.1706[47]	0.1381[37]
195	yoonik-000	0.0520[165]	0.1056[86]	0.2675[91]	0.3773[61]	0.0895[48]	0.1823[48]	0.3214[38]	0.1027[52]	0.2582[56]	0.3628[43]	-	-
196	yoonik-001	0.0287[148]	0.0507[59]	0.1464[62]	0.2009[43]	-	-	-	0.0526[43]	0.1508[41]	-	-	-
197	yuan-000	0.0130[118]	0.0871[76]	0.2704[95]	0.3516[56]	0.0654[43]	0.1661[46]	-	0.0805[47]	0.2492[55]	0.3645[44]	-	-
198	yuan-001	0.0083[81]	-	0.3019[103]	-	-	-	-	-	-	-	-	-

Table 10: This table summarizes False Non-Match Rate (FNMR) on unmasked and masked probe images. FNMR is the proportion of mated comparisons below a threshold set to achieve FMR=1e-05 on unmasked probe images. False Match Rate (FMR) is the proportion of impostor comparisons at or above that threshold. The red superscripts give rank over all algorithms in that column. Missing entries generally mean the algorithm was not run on that particular mask variation due to time and resource constraints. Algorithms with FTE=1.00 were not run at all. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 11: At a fixed threshold, a plot of FNMR with and without masks. The displacement of the dark red line relative to the black "parity" line shows a large increase in FNMR with masks for pre-COVID algorithms. The reduction in distance (relative to the black line) observed in the gold line indicates a reduction in median FNMR with masks for post-COVID algorithms. The value in the title is the median increase multiplier.

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Figure 12: At a fixed threshold, a plot of FNMR with round versus wide masks. The displacement of the dark red (pre-COVID algorithms) and gold (post-COVID algorithms) lines relative to the black "parity" line shows a modest increase in FNMR with wide masks, with median post-COVID results showing nominal FNMR reductions. The value in the title is the median increase multiplier.

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Figure 13: At a fixed threshold, a plot of FNMR with low versus high masks. The displacement of the dark red (pre-COVID algorithms) and gold (post-COVID algorithms) lines relative to the black "parity" lines shows a considerable increase in FNMR with high vs. low nose coverage masks, with median post-COVID results showing modest FNMR reductions. The value in the title is the median increase multiplier.

	Algorithm Name	COLOR = WHITE SHAPE = WIDE			COLOR = LIGHTBLUE SHAPE = WIDE			COLOR = LIGHTBLUE SHAPE = ROUND			COLOR = RED SHAPE = WIDE			COLOR = BLACK SHAPE = WIDE		
	COVERAGE	LO	MED	HI	LO	MED	HI	LO	MED	HI	LO	MED	HI	LO	MED	HI
1	3divi-004	0.514	0.659	0.627	0.420	0.599	0.603	0.378	0.663	0.769	0.290	0.691	0.864	0.653	0.920	0.939
2	3divi-005	0.003	0.012	0.029	0.002	0.013	0.030	0.007	0.011	0.026	0.005	0.040	0.077	0.002	0.006	0.012
3	acer-000	0.048	0.105	0.139	0.035	0.080	0.114	0.052	0.078	0.137	0.168	0.321	0.445	0.107	0.197	0.270
4	acer-001	0.020	0.048	0.091	0.019	0.051	0.096	0.025	0.034	0.060	0.025	0.080	0.147	0.026	0.088	0.157
5	acisw-003	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
6	acisw-006	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
7	adera-002	0.103	0.167	0.205	0.094	0.141	0.193	0.098	0.116	0.180	0.267	0.457	0.551	0.214	0.333	0.397
8	advance-002	0.019	0.046	0.096	0.020	0.045	0.096	0.026	0.037	0.085	0.028	0.088	0.174	0.034	0.104	0.200
9	aifirst-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
10	aigen-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
11	ailabs-001	0.071	0.208	0.248	0.061	0.194	0.233	0.102	0.177	0.314	0.096	0.319	0.389	0.116	0.310	0.465
12	aimall-002	0.073	0.129	0.225	0.095	0.152	0.260	0.107	0.159	0.236	0.086	0.122	0.230	0.049	0.071	0.154
13	aimall-003	0.011	0.038	0.097	0.014	0.047	0.113	0.020	0.041	0.085	0.013	0.039	0.103	0.012	0.033	0.081
14	aiunionface-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
15	alleyes-000	0.006	0.023	0.062	0.006	0.020	0.056	0.007	0.012	0.028	0.007	0.035	0.088	0.010	0.043	0.104
16	alphaface-002	0.025	0.056	0.099	0.024	0.054	0.095	0.033	0.044	0.072	0.041	0.104	0.174	0.027	0.071	0.132
17	androvideo-000	-	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
18	anke-005	0.009	0.028	0.066	0.011	0.030	0.069	0.012	0.018	0.041	0.013	0.036	0.079	0.009	0.056	0.091
19	antheus-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
20	antheus-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
21	anyvision-005	0.003	0.012	0.029	0.003	0.032	0.069	0.008	0.018	0.045	0.007	0.045	0.123	0.004	0.046	0.137
22	asusaics-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
23	aware-005	0.053	0.151	0.218	0.039	0.129	0.211	0.046	0.089	0.244	0.095	0.380	0.516	0.091	0.236	0.449
24	awiros-001	0.195	0.370	0.450	0.162	0.298	0.379	0.161	0.258	0.355	0.388	0.650	0.772	0.198	0.415	0.642
25	awiros-002	0.101	0.212	0.215	0.103	0.202	0.220	0.187	0.261	0.356	0.071	0.272	0.296	0.232	0.405	0.455
26	ayftech-001	0.237	0.610	0.587	0.251	0.615	0.609	0.301	0.553	0.728	0.105	0.405	0.626	0.193	0.596	0.758
27	bioidtechswiss-001	0.004	0.014	0.044	0.004	0.017	0.046	0.006	0.011	0.026	0.005	0.020	0.051	0.007	0.022	0.056
28	bioidtechswiss-002	0.004	0.013	0.041	0.003	0.017	0.045	0.006	0.011	0.026	0.004	0.018	0.049	0.004	0.017	0.047
29	bresee-000	0.005	0.022	0.049	0.005	0.020	0.049	0.009	0.014	0.034	0.010	0.051	0.108	0.008	0.041	0.097
30	bresee-001	0.005	0.022	0.049	0.005	0.020	0.049	0.009	0.014	0.034	0.010	0.051	0.108	0.008	0.041	0.097
31	camvi-004	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
32	canon-002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
33	ceiec-003	0.023	0.056	0.103	0.021	0.054	0.096	0.029	0.042	0.100	0.048	0.143	0.218	0.044	0.128	0.222
34	ceiec-004	0.008	0.021	0.047	0.008	0.021	0.045	0.010	0.015	0.033	0.012	0.042	0.094	0.015	0.051	0.099
35	chosun-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
36	chosun-002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
37	chtface-002	0.033	0.100	0.154	0.026	0.081	0.126	0.031	0.056	0.107	0.110	0.321	0.480	0.042	0.144	0.270
38	chtface-003	0.022	0.075	0.125	0.018	0.059	0.099	0.023	0.039	0.077	0.080	0.259	0.418	0.030	0.108	0.224
39	cib-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
40	clova-000	0.035	0.089	0.149	0.033	0.081	0.136	0.039	0.059	0.128	0.071	0.245	0.361	0.049	0.146	0.252
41	cogent-005	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
42	cognitec-002	0.136	0.142	0.196	0.160	0.247	0.298	0.221	0.326	0.390	0.115	0.299	0.518	0.144	0.446	0.721
43	cor-001	0.002	0.010	0.031	0.003	0.013	0.035	0.004	0.008	0.020	0.004	0.013	0.032	0.004	0.011	0.031
44	csc-001	0.035	0.100	0.180	0.040	0.121	0.213	0.088	0.123	0.317	0.044	0.123	0.241	0.054	0.154	0.308
45	ctbcbank-000	0.179	0.794	0.803	0.171	0.786	0.865	0.205	0.620	0.915	0.433	0.857	0.913	0.189	0.806	0.895

Table 11: This table summarizes Failure to Enroll (FTE) rates surveyed over 10 000 images of each mask variant. FTE is the proportion of failed template generation attempts. Failures can occur because the software throws an exception, or because the software electively refuses to process the input image as would occur if the algorithms does not detect a face or determines that the face has insufficient information. FTE is measured as the number of function calls that give EITHER a non-zero error code OR that give a "small" template containing fewer than 60 bytes. This second rule is needed because some algorithms incorrectly fail to return a non-zero error code when template generation fails but do produce a skeletal template. The effects of FTE are included in the accuracy results of this report by regarding any template comparison involving a failed template to produce a low similarity score. Thus higher FTE results in higher FNMR and lower FMR. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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	Algorithm Name	COLOR = WHITE SHAPE = WIDE			COLOR = LIGHTBLUE SHAPE = WIDE			COLOR = LIGHTBLUE SHAPE = ROUND			COLOR = RED SHAPE = WIDE			COLOR = BLACK SHAPE = WIDE		
	COVERAGE	LO	MED	HI	LO	MED	HI	LO	MED	HI	LO	MED	HI	LO	MED	HI
46	cubox-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
47	cuhkee-001	0.009	0.029	0.069	0.009	0.031	0.074	0.014	0.025	0.057	0.012	0.048	0.115	0.013	0.048	0.140
48	cybercore-000	0.194	0.376	0.371	0.222	0.447	0.453	0.267	0.461	0.751	0.191	0.432	0.554	0.367	0.678	0.805
49	cyberlink-005	0.012	0.035	0.080	0.011	0.029	0.067	0.014	0.021	0.043	0.016	0.072	0.147	0.015	0.049	0.114
50	cyberlink-006	0.002	0.007	0.016	0.001	0.006	0.015	0.001	0.002	0.003	0.003	0.011	0.030	0.002	0.011	0.029
51	dahua-005	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
52	dahua-006	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
53	decatur-000	0.014	0.036	0.074	0.014	0.033	0.069	0.018	0.028	0.058	0.032	0.065	0.135	0.033	0.069	0.135
54	deepglint-002	0.002	0.009	0.028	0.002	0.012	0.031	0.004	0.006	0.017	0.003	0.014	0.035	0.003	0.010	0.024
55	deepsea-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
56	dermalog-007	0.013	0.041	0.064	0.015	0.045	0.072	0.018	0.027	0.049	0.013	0.042	0.067	0.016	0.056	0.079
57	didiglobalface-001	0.025	0.056	0.099	0.024	0.054	0.095	0.033	0.044	0.072	0.041	0.104	0.174	0.027	0.071	0.132
58	dsk-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
59	eocortex-000	0.692	0.862	0.861	0.653	0.841	0.860	0.647	0.887	0.904	0.800	0.960	0.990	0.814	0.950	0.990
60	ercacat-001	0.002	0.007	0.014	0.002	0.006	0.013	0.002	0.002	0.006	0.002	0.008	0.016	0.003	0.009	0.017
61	expasoft-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
62	expasoft-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
63	f8-001	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
64	facesoft-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
65	fujitsulab-001	0.008	0.014	0.032	0.006	0.011	0.020	0.005	0.005	0.008	0.014	0.038	0.108	0.012	0.025	0.066
66	fujitsulab-002	0.005	0.011	0.026	0.004	0.009	0.017	0.004	0.004	0.007	0.012	0.035	0.095	0.009	0.022	0.058
67	geo-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
68	geo-001	-	-	-	0.012	0.027	0.070	-	-	-	-	-	-	0.016	0.052	-
69	glory-002	0.059	0.106	0.128	0.056	0.101	0.124	0.053	0.074	0.126	0.062	0.148	0.239	0.054	0.154	0.279
70	glory-003	0.042	0.115	0.157	0.039	0.097	0.143	0.051	0.068	0.107	0.061	0.168	0.234	0.048	0.196	0.272
71	gorilla-005	0.006	0.018	0.040	0.007	0.018	0.038	0.009	0.012	0.024	-	-	-	0.012	0.037	0.071
72	gorilla-006	0.003	0.008	0.014	0.003	0.008	0.013	0.002	0.003	0.005	0.005	0.013	0.034	0.005	0.014	0.031
73	hertasecurity-000	0.022	0.083	0.188	0.022	0.083	0.192	0.029	0.046	0.161	0.023	0.095	0.212	0.028	0.144	0.277
74	hr-003	0.003	0.008	0.013	0.003	0.009	0.012	0.006	0.008	0.014	0.005	0.011	0.022	0.005	0.009	0.017
75	hr-004	0.004	0.010	0.015	0.005	0.009	0.013	0.004	0.005	0.007	0.005	0.013	0.018	0.005	0.011	0.019
76	hv-001	0.082	0.202	0.290	0.069	0.155	0.227	0.073	0.124	0.207	0.227	0.477	0.662	0.084	0.239	0.484
77	icm-002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
78	icthtc-000	0.114	0.272	0.362	0.095	0.208	0.286	0.101	0.176	0.292	0.293	0.586	0.758	0.118	0.320	0.581
79	id3-005	0.143	0.256	0.304	0.169	0.364	0.424	0.196	0.287	0.539	0.192	0.395	0.527	0.216	0.389	0.557
80	id3-006	0.015	0.051	0.123	0.024	0.068	0.150	0.040	0.081	0.157	0.015	0.050	0.124	0.019	0.059	0.134
81	idemia-006	0.002	0.008	0.028	0.002	0.007	0.023	0.002	0.004	0.015	0.002	0.013	0.035	0.002	0.010	0.029
82	idemia-007	0.002	0.008	0.027	0.002	0.006	0.023	0.002	0.003	0.014	0.002	0.013	0.035	0.002	0.009	0.028
83	iit-002	0.012	0.036	0.074	0.013	0.043	0.091	0.015	0.027	0.072	0.016	0.060	0.137	0.015	0.087	0.185
84	iit-003	0.004	0.018	0.050	0.005	0.021	0.055	0.006	0.013	0.039	0.005	0.020	0.052	0.005	0.022	0.053
85	imagus-001	0.016	0.040	0.074	0.014	0.037	0.066	0.023	0.029	0.056	0.023	0.084	0.135	0.021	0.085	0.149
86	imagus-002	0.004	0.013	0.034	0.004	0.012	0.030	0.005	0.007	0.016	0.006	0.019	0.052	0.006	0.021	0.056
87	imperial-002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
88	incode-007	0.003	0.013	0.034	0.003	0.013	0.031	0.003	0.006	0.014	0.004	0.016	0.045	0.004	0.020	0.060
89	incode-008	0.003	0.013	0.034	0.003	0.013	0.031	0.003	0.006	0.014	0.004	0.016	0.045	0.004	0.020	0.060
90	innefulabs-000	0.007	0.024	0.064	0.009	0.028	0.069	0.014	0.027	0.053	0.008	0.025	0.060	0.009	0.025	0.064

Table 12: This table summarizes Failure to Enroll (FTE) rates surveyed over 10 000 images of each mask variant. FTE is the proportion of failed template generation attempts. Failures can occur because the software throws an exception, or because the software electively refuses to process the input image as would occur if the algorithms does not detect a face or determines that the face has insufficient information. FTE is measured as the number of function calls that give EITHER a non-zero error code OR that give a "small" template containing fewer than 60 bytes. This second rule is needed because some algorithms incorrectly fail to return a non-zero error code when template generation fails but do produce a skeletal template. The effects of FTE are included in the accuracy results of this report by regarding any template comparison involving a failed template to produce a low similarity score. Thus higher FTE results in higher FNMR and lower FMR. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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	Algorithm Name	COLOR = WHITE SHAPE = WIDE			COLOR = LIGHTBLUE SHAPE = WIDE			COLOR = LIGHTBLUE SHAPE = ROUND			COLOR = RED SHAPE = WIDE			COLOR = BLACK SHAPE = WIDE		
	COVERAGE	LO	MED	HI	LO	MED	HI	LO	MED	HI	LO	MED	HI	LO	MED	HI
91	innovativetechnologyltd-002	0.082	0.176	0.232	0.074	0.172	0.233	0.091	0.131	0.265	0.132	0.269	0.391	0.149	0.362	0.516
92	innovatrics-006	0.002	0.017	0.051	0.003	0.018	0.054	0.005	0.012	0.035	0.007	0.045	0.103	0.005	0.037	0.087
93	innovatrics-007	0.001	0.005	0.015	0.002	0.005	0.017	0.003	0.003	0.007	0.003	0.012	0.038	0.002	0.009	0.026
94	intellicloudai-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
95	intellicloudai-002	0.007	0.015	0.028	0.005	0.013	0.025	0.007	0.009	0.015	0.012	0.041	0.087	0.013	0.031	0.056
96	intellifusion-002	0.000	0.001	0.004	0.000	0.000	0.001	0.000	0.000	0.002	0.001	0.035	0.112	0.000	0.001	0.004
97	intellivision-002	0.073	0.213	0.267	0.068	0.210	0.261	0.143	0.204	0.340	0.308	0.517	0.618	0.137	0.396	0.469
98	intelresearch-002	0.006	0.016	0.031	0.007	0.017	0.032	0.008	0.011	0.025	0.008	0.021	0.046	0.008	0.023	0.052
99	intelresearch-003	0.002	0.008	0.020	0.002	0.007	0.018	0.003	0.004	0.009	0.003	0.010	0.031	0.003	0.009	0.021
100	intsysmsu-002	0.008	0.055	0.117	0.007	0.047	0.110	0.015	0.033	0.100	0.010	0.078	0.191	0.036	0.105	0.231
101	iqface-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
102	iqface-003	0.049	0.079	0.124	0.047	0.080	0.127	0.049	0.067	0.114	0.058	0.155	0.227	0.053	0.148	0.219
103	irex-000	0.003	0.011	0.027	0.003	0.012	0.026	0.004	0.006	0.012	0.003	0.020	0.054	0.006	0.032	0.066
104	isap-001	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
105	isap-002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
106	itmo-007	0.008	0.034	0.086	0.009	0.046	0.106	0.017	0.034	0.071	0.009	0.039	0.102	0.011	0.034	0.082
107	ivacognitive-001	0.011	0.027	0.072	0.011	0.027	0.070	0.015	0.021	0.051	0.015	0.040	0.123	0.016	0.049	0.146
108	iws-000	0.489	0.829	0.795	0.476	0.827	0.804	0.627	0.800	0.937	0.387	0.873	0.895	0.512	0.961	0.984
109	kakao-003	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
110	kakao-004	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
111	kedacom-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
112	kneron-005	0.063	0.184	0.206	0.058	0.166	0.212	0.094	0.146	0.276	0.080	0.292	0.309	0.101	0.440	0.505
113	kookmin-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
114	lookman-004	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
115	luxand-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
116	minivision-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
117	mobbl-000	0.610	0.654	0.599	0.512	0.553	0.539	0.547	0.630	0.605	0.569	0.771	0.868	0.849	0.915	0.936
118	moredian-000	0.008	0.031	0.085	0.011	0.037	0.091	0.018	0.037	0.069	0.009	0.030	0.078	0.010	0.030	0.079
119	mt-002	0.001	0.004	0.007	0.001	0.004	0.006	0.001	0.001	0.003	0.001	0.005	0.013	0.002	0.005	0.009
120	mt-003	0.003	0.015	0.046	0.004	0.017	0.049	0.007	0.013	0.031	0.005	0.018	0.045	0.006	0.017	0.044
121	mvision-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
122	nazhiai-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
123	netbridgetech-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
124	netbridgetech-002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
125	neurotechnology-009	0.006	0.018	0.026	0.005	0.015	0.023	0.004	0.005	0.014	0.009	0.030	0.043	0.015	0.100	0.086
126	neurotechnology-010	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.001	0.000	0.001	0.000	0.001	0.001
127	nodeflux-002	0.402	0.598	0.538	0.440	0.671	0.628	0.482	0.681	0.877	0.393	0.672	0.772	0.602	0.835	0.915
128	notiontag-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
129	nsensecorp-001	0.074	0.139	0.196	0.089	0.174	0.241	0.073	0.119	0.264	0.073	0.137	0.204	0.083	0.219	0.291
130	ntechlab-008	0.064	0.126	0.196	0.053	0.011	0.183	0.003	0.095	0.018	0.003	0.016	0.043	0.003	0.016	0.042
131	ntechlab-009	0.011	0.023	0.040	0.011	0.022	0.040	0.013	0.013	0.023	0.012	0.030	0.047	0.013	0.027	0.044
132	oz-001	0.010	0.030	0.057	0.010	0.034	0.066	0.024	0.031	0.063	0.036	0.164	0.209	0.007	0.027	0.040
133	oz-002	0.001	0.002	0.005	0.001	0.002	0.004	0.001	0.001	0.002	0.001	0.002	0.005	0.001	0.002	0.006
134	paravision-004	0.002	0.011	0.027	0.002	0.010	0.024	0.003	0.004	0.009	0.003	0.019	0.056	0.003	0.016	0.043
135	paravision-006	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000

Table 13: This table summarizes Failure to Enroll (FTE) rates surveyed over 10 000 images of each mask variant. FTE is the proportion of failed template generation attempts. Failures can occur because the software throws an exception, or because the software electively refuses to process the input image as would occur if the algorithms does not detect a face or determines that the face has insufficient information. FTE is measured as the number of function calls that give EITHER a non-zero error code OR that give a "small" template containing fewer than 60 bytes. This second rule is needed because some algorithms incorrectly fail to return a non-zero error code when template generation fails but do produce a skeletal template. The effects of FTE are included in the accuracy results of this report by regarding any template comparison involving a failed template to produce a low similarity score. Thus higher FTE results in higher FNMR and lower FMR. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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	Algorithm Name	COLOR = WHITE SHAPE = WIDE			COLOR = LIGHTBLUE SHAPE = WIDE			COLOR = LIGHTBLUE SHAPE = ROUND			COLOR = RED SHAPE = WIDE			COLOR = BLACK SHAPE = WIDE		
	COVERAGE	LO	MED	HI	LO	MED	HI	LO	MED	HI	LO	MED	HI	LO	MED	HI
136	pensees-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
137	pixelall-003	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
138	pixelall-004	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
139	pixelall-005	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
140	psl-005	0.004	0.015	0.038	0.003	0.014	0.036	0.006	0.011	0.029	0.006	0.025	0.061	0.008	0.020	0.047
141	psl-006	0.000	0.001	0.002	0.000	0.001	0.002	0.001	0.001	0.002	0.000	0.002	0.004	0.000	0.001	0.004
142	ptakuratsatu-000	0.001	0.005	0.015	0.002	0.005	0.017	0.003	0.003	0.007	0.003	0.012	0.038	0.002	0.009	0.026
143	pxl-001	0.096	0.268	0.329	0.082	0.208	0.273	0.090	0.161	0.294	0.272	0.579	0.712	0.116	0.302	0.500
144	rankone-009	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.001	0.000	0.000	0.001
145	rankone-010	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
146	remarkai-002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
147	rokid-000	0.194	0.372	0.370	0.220	0.444	0.450	0.265	0.457	0.749	0.191	0.431	0.553	0.367	0.677	0.806
148	s1-001	0.647	0.943	0.911	0.617	0.930	0.915	0.616	0.919	0.954	0.613	0.957	0.960	0.646	0.962	0.962
149	scanovate-001	0.544	0.601	0.596	0.515	0.553	0.579	0.513	0.565	0.664	0.766	0.914	0.960	0.554	0.676	0.806
150	scanovate-002	0.008	0.019	0.046	0.008	0.018	0.045	0.008	0.012	0.027	0.010	0.025	0.079	0.012	0.036	0.101
151	securifai-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
152	sensetime-003	0.009	0.029	0.069	0.009	0.031	0.074	0.014	0.025	0.057	0.012	0.048	0.115	0.013	0.048	0.140
153	sensetime-004	0.009	0.027	0.066	0.009	0.029	0.070	0.013	0.023	0.053	0.012	0.044	0.111	0.012	0.045	0.135
154	sertis-000	0.002	0.012	0.034	0.002	0.012	0.032	0.003	0.005	0.013	0.003	0.016	0.047	0.005	0.020	0.052
155	sertis-001	0.002	0.006	0.017	0.002	0.006	0.014	0.001	0.002	0.005	0.002	0.007	0.017	0.003	0.009	0.020
156	shu-002	0.011	0.031	0.080	0.009	0.026	0.083	0.023	0.037	0.103	0.026	0.126	0.227	0.016	0.056	0.167
157	shu-003	0.003	0.006	0.013	0.002	0.004	0.008	0.003	0.005	0.009	0.008	0.031	0.070	0.004	0.014	0.040
158	sjtu-002	0.011	0.031	0.080	0.009	0.026	0.083	0.023	0.037	0.103	0.026	0.126	0.227	0.016	0.056	0.167
159	sjtu-003	0.002	0.004	0.009	0.002	0.003	0.006	0.002	0.003	0.006	0.005	0.022	0.049	0.003	0.009	0.028
160	staqu-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
161	starhybrid-001	0.192	0.468	0.461	0.149	0.406	0.483	0.137	0.321	0.487	0.478	0.761	0.781	0.133	0.372	0.565
162	synesis-006	0.001	0.003	0.007	0.001	0.003	0.007	0.001	0.001	0.003	0.001	0.005	0.011	0.001	0.004	0.008
163	synesis-007	0.007	0.021	0.047	0.008	0.028	0.056	0.013	0.025	0.056	0.007	0.046	0.114	0.018	0.124	0.134
164	synology-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
165	synology-002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
166	sztu-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
167	tech5-004	0.005	0.022	0.061	0.006	0.028	0.070	0.010	0.021	0.046	0.006	0.025	0.068	0.006	0.021	0.058
168	tech5-005	0.004	0.015	0.044	0.004	0.017	0.046	0.006	0.012	0.026	0.005	0.020	0.051	0.007	0.022	0.057
169	tevian-005	0.125	0.463	0.370	0.148	0.650	0.557	0.208	0.359	0.705	0.086	0.667	0.679	0.131	0.786	0.787
170	tevian-006	0.012	0.089	0.160	0.010	0.066	0.124	0.028	0.041	0.146	0.034	0.266	0.408	0.016	0.256	0.307
171	tiger-004	0.015	0.037	0.075	0.014	0.034	0.070	0.018	0.029	0.060	0.033	0.070	0.144	0.034	0.069	0.138
172	trueface-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
173	trueface-001	0.078	0.163	0.217	0.071	0.153	0.217	0.084	0.119	0.268	-	-	-	0.147	0.317	0.474
174	tuputech-000	0.517	0.679	0.684	0.626	0.758	0.765	0.502	0.619	0.714	0.652	0.893	0.926	0.661	0.904	0.933
175	uluface-002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
176	upc-001	0.002	0.005	0.012	0.002	0.005	0.012	0.002	0.002	0.005	0.002	0.010	0.027	0.003	0.007	0.018
177	veridas-004	0.047	0.139	0.202	0.041	0.112	0.164	0.048	0.085	0.154	0.106	0.401	0.568	0.054	0.213	0.383
178	veridas-005	0.047	0.136	0.200	0.041	0.110	0.161	0.046	0.083	0.152	0.106	0.401	0.568	0.053	0.208	0.382
179	via-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
180	videmo-000	0.019	0.067	0.125	0.018	0.051	0.106	0.023	0.040	0.089	0.122	0.352	0.368	0.027	0.100	0.296

Table 14: This table summarizes Failure to Enroll (FTE) rates surveyed over 10 000 images of each mask variant. FTE is the proportion of failed template generation attempts. Failures can occur because the software throws an exception, or because the software electively refuses to process the input image as would occur if the algorithms does not detect a face or determines that the face has insufficient information. FTE is measured as the number of function calls that give EITHER a non-zero error code OR that give a "small" template containing fewer than 60 bytes. This second rule is needed because some algorithms incorrectly fail to return a non-zero error code when template generation fails but do produce a skeletal template. The effects of FTE are included in the accuracy results of this report by regarding any template comparison involving a failed template to produce a low similarity score. Thus higher FTE results in higher FNMR and lower FMR. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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	Algorithm	COLOR = WHITE			COLOR = LIGHTBLUE						COLOR = RED			COLOR = BLACK		
	Name	SHAPE = WIDE			SHAPE = WIDE			SHAPE = ROUND			SHAPE = WIDE			SHAPE = WIDE		
	COVERAGE	LO	MED	HI	LO	MED	HI	LO	MED	HI	LO	MED	HI	LO	MED	HI
181	videonetics-002	0.338	0.581	0.557	0.330	0.569	0.542	0.378	0.559	0.785	0.344	0.693	0.735	0.396	0.702	0.848
182	vigilantsolutions-008	0.062	0.168	0.218	0.051	0.136	0.189	0.068	0.124	0.202	0.096	0.407	0.633	0.072	0.268	0.492
183	vigilantsolutions-009	0.062	0.168	0.218	0.051	0.136	0.189	0.068	0.124	0.202	0.096	0.407	0.633	0.072	0.268	0.492
184	vinai-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
185	visage-000	0.165	0.548	0.544	0.152	0.508	0.505	0.173	0.379	0.729	0.171	0.667	0.672	0.245	0.776	0.807
186	visionlabs-009	0.011	0.037	0.082	0.010	0.035	0.076	0.018	0.042	0.097	0.016	0.052	0.120	0.024	0.065	0.150
187	visionlabs-010	0.002	0.004	0.007	0.001	0.003	0.004	0.001	0.001	0.001	0.001	0.002	0.003	0.001	0.003	0.004
188	visteam-000	0.058	0.150	0.210	0.048	0.118	0.176	0.052	0.092	0.156	0.160	0.372	0.506	0.074	0.202	0.369
189	vnpt-001	0.914	0.967	0.967	0.918	0.970	0.974	0.894	0.968	0.974	0.801	0.896	0.916	0.883	0.956	0.972
190	vocord-008	0.013	0.046	0.087	0.011	0.052	0.089	0.031	0.059	0.111	0.013	0.079	0.127	0.009	0.050	0.093
191	vocord-009	0.002	0.008	0.018	0.002	0.006	0.013	0.002	0.003	0.006	0.002	0.009	0.020	0.002	0.010	0.026
192	vts-000	0.011	0.029	0.072	0.012	0.028	0.074	0.015	0.020	0.048	0.014	0.043	0.129	0.016	0.051	0.149
193	winsense-001	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
194	winsense-002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
195	xforwardai-001	0.006	0.025	0.069	0.007	0.031	0.079	0.011	0.021	0.052	0.007	0.030	0.079	0.007	0.022	0.060
196	xforwardai-002	0.006	0.025	0.069	0.007	0.031	0.079	0.011	0.021	0.052	0.007	0.030	0.079	0.007	0.022	0.060
197	xm-000	0.003	0.006	0.013	0.002	0.004	0.008	0.003	0.004	0.009	0.008	0.031	0.070	0.005	0.013	0.038
198	yoonik-000	0.029	0.061	0.138	0.027	0.060	0.142	0.037	0.049	0.124	0.040	0.094	0.219	0.039	0.099	0.257
199	yoonik-001	0.015	0.052	0.110	0.017	0.055	0.113	0.024	0.035	0.072	0.031	0.093	0.170	0.026	0.086	0.175
200	yuan-000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
201	yuan-001	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000

Table 15: This table summarizes Failure to Enroll (FTE) rates surveyed over 10 000 images of each mask variant. FTE is the proportion of failed template generation attempts. Failures can occur because the software throws an exception, or because the software electively refuses to process the input image as would occur if the algorithms does not detect a face or determines that the face has insufficient information. FTE is measured as the number of function calls that give EITHER a non-zero error code OR that give a "small" template containing fewer than 60 bytes. This second rule is needed because some algorithms incorrectly fail to return a non-zero error code when template generation fails but do produce a skeletal template. The effects of FTE are included in the accuracy results of this report by regarding any template comparison involving a failed template to produce a low similarity score. Thus higher FTE results in higher FNMR and lower FMR. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 14: For each algorithm the rightmost dot shows FNMR @ FMR=0.00001 (as reported throughout this report). The left most dot shows the failure-to-template (FTE) rate over the masked verification set of 5.2M images. The gap between the two dots is attributable to low similarity score. Some FTE rates are zero - rates below 0.001 are shown as 0.001.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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The following plots are detection error tradeoff (DET) characteristics for each algorithm, across different mask nose coverages and shapes.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 15: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 16: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 17: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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FMR(T) "False match rate"

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Figure 18: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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FMR(T) "False match rate"

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Figure 19: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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FMR(T) "False match rate"

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Figure 20: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 21: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 22: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 23: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 24: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 25: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 26: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 27: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 28: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 29: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 30: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 31: DET curves showing error rates on unmasked and masked probe images, broken out by mask shape and nose coverage. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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The following plots are detection error tradeoff (DET) characteristics for each algorithm, across different mask colors.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 32: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 33: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 34: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 35: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 36: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 37: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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FMR(T) "False match rate"

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Figure 38: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 39: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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FMR(T) "False match rate"

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Figure 40: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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FMR(T) "False match rate"

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Figure 41: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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FMR(T) "False match rate"

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Figure 42: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 43: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 44: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 45: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 46: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 47: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 48: DET curves showing error rates on unmasked and masked probe images, broken out by mask color. Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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The following plots show the explicit dependence of false non-match rate (FNMR) and false match rate (FMR) on score threshold for each algorithm, across different masked/unmasked combinations.

FNMR(T) "False non-match rate"
FMR(T) "False match rate"

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Figure 49: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 50: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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FMR(T) "False match rate"

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Figure 51: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 52: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 53: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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FMR(T) "False match rate"

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Figure 54: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 55: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 56: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 57: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 58: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 59: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 60: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 61: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 62: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 63: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 64: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 65: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 66: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 67: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 68: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 69: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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Figure 70: FNMR and FMR calibration curves on unmasked-to-unmasked (both unmasked), masked-to-unmasked with medium, wide, lightblue masks (one masked), masked-to-masked with medium, wide, lightblue masks (both masked (w-w)), and masked-to-masked with medium, round, white masks (enrollment image) and medium, wide, lightblue masks (verification image) (both masked (r-w)). Algorithms in black were submitted prior to mid-March 2020, and algorithms in blue were submitted thereafter.

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References

[1] ISO/IEC 19794-5:2011 - Information technology - Biometric data interchange formats - Part 5: Face image data.

[2] N95 Respirators, Surgical Masks, and Face Masks. https://www.fda.gov/medical-devices/personal-protective-equipment-infection-control/n95-respirators-surgical-masks-and-face-masks.

[3] NIST Special Database 32 - Multiple Encounter Dataset (MEDS-II). https://www.nist.gov/itl/iad/image-group/special-database-32-multiple-encounter-dataset-meds.

[4] Patrick Grother, Mei Ngan, and Kayee Hanaoka. NISTIR 8311 - Ongoing FRVT Part 6A: Face recognition accuracy with face masks using pre-COVID-19 algorithms. Interagency Report 8311, National Institute of Standards and Technology, July 2017. https://pages.nist.gov/frvt/reports/11/frvt_11_report_2017_07_29.pdf.

[5] Patrick Grother, George Quinn, and Mei Ngan. Face in video evaluation (five) face recognition of non-cooperative subjects. Interagency Report 8173, National Institute of Standards and Technology, March 2017. https://doi.org/10.6028/NIST.IR.8173.

[6] J. Hanley and A. Lippman-Hand. If nothing goes wrong, is everything all right? interpreting zero numerators. *JAMA*, 249 13:1743–5, 1983.

[7] Davis E. King. Dlib-ml: A machine learning toolkit. *Journal of Machine Learning Research*, 10:1755–1758, 2009. http://dlib.net.

[8] Mei Ngan, Patrick Grother, and Kayee Hanaoka. NISTIR 8311 - Ongoing FRVT Part 6A: Face recognition accuracy with face masks using pre-COVID-19 algorithms. Interagency Report 8311, National Institute of Standards and Technology, July 2020. https://doi.org/10.6028/NIST.IR.8311.

[9] Patrick Grother and Mei Ngan and Kayee Hanaoka. NIST Ongoing Face Recognition Vendor Test (FRVT) 1:1 Verification Application Programming Interface (API), April 2019. https://pages.nist.gov/frvt/api/FRVT_ongoing_11_api_4.0.pdf.

[10] Zhongyuan Wang, Guangcheng Wang, Baojin Huang, Zhangyang Xiong, Qi Hong, Hao Wu, Peng Yi, Kui Jiang, Nanxi Wang, Yingjiao Pei, Heling Chen, Yu Miao, Zhibing Huang, and Jinbi Liang. Masked face recognition dataset and application, 2020.

Appendix A Dlib Masking Methodology

(a) wide, high coverage

(b) round, high coverage

(c) wide, medium coverage

(d) round, medium coverage

(e) wide, low coverage

(f) round, low coverage

Figure 71: This figure shows the Dlib facial points used to create the various synthetic masks used in this report. For wide masks, the specified Dlib facial points were used to generate a closed polygon and two additional points were interpolated between each dlib facial point used for smoothing purposes. For round masks, the specified Dlib facial points were used to generate an ellipse. The Dlib C++ toolkit version 19.19, configured with the common histogram of gradients (HoG)-based face detector and 68 face landmark shape predictor was used to generate the 68 facial landmarks.

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