

Open Media Forensics Challenge (OpenMFC) 2021 Workshop

OpenMFC Introduction

Yooyoung Lee, Haiying Guan, Lukas Diduch, and Ilia Ghorbanian

> Multimodal Information Group, Information Access Division

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• NIST contributors

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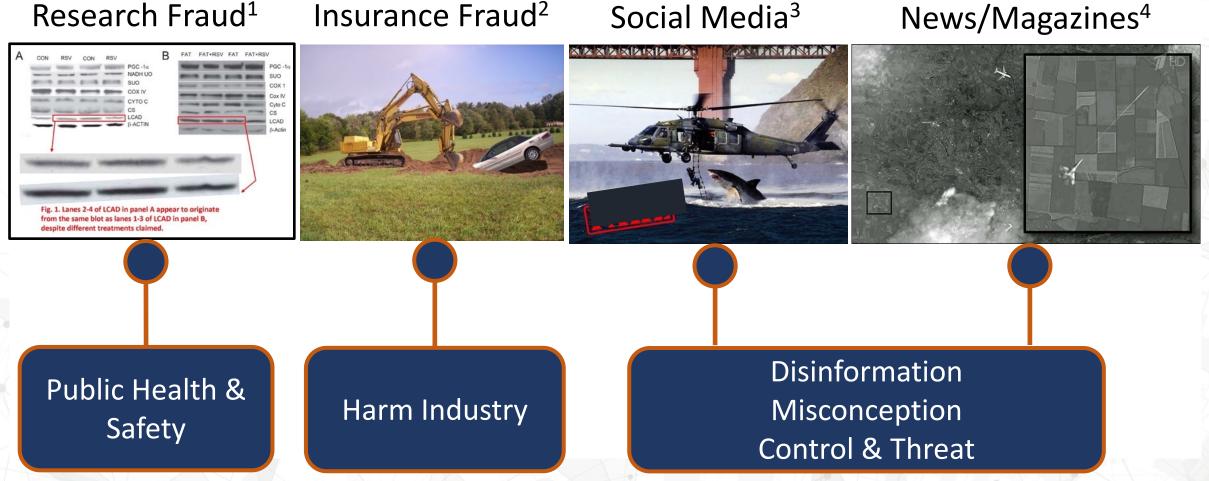
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Motivation

Media Forensics is an attempt to determine the authenticity of digital media



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MediFor: Media Forensics (2017 - 2020)



Sponsor: DARPA MediFor program (PM: Matt Turek)

Definition: determine the authenticity and establish the integrity of visual/audio media



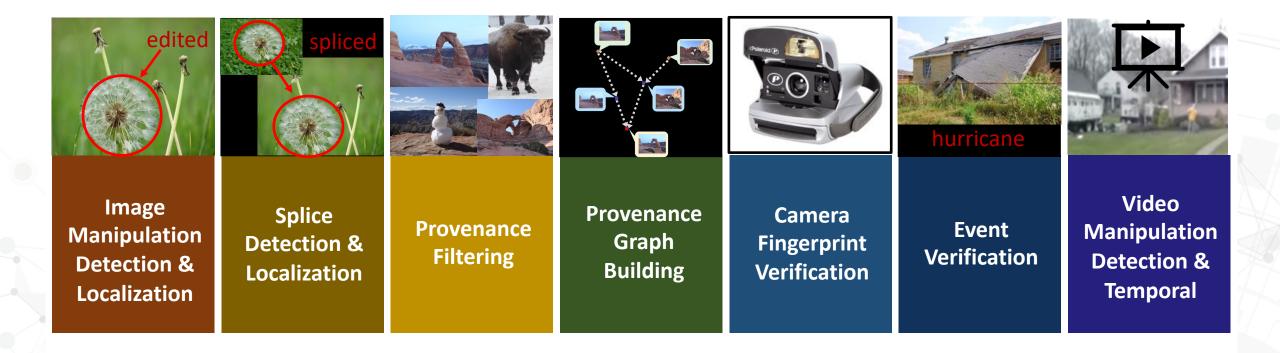
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Objective: develop technologies to advance the field of forensics



NIST role: define tasks and metrics, and manage technical evaluations of media forensic technologies

MediFor Evaluation Tasks



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MediFor at A Glance





Why Was It Challenging?



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- Large variety of disciplines or domains
- **Evaluation design challenges**
- Lack of benchmark datasets
- Different data collections and annotations



Complex scoring protocolsHolistic, Opt-In, Selective, and special studies

High complexity

Large effort and time

Broad scope



Multiple evaluation infrastructures

• Open (take-home) vs Container (sequester)

> Participation difficulty



Goal: automatically detect and locate manipulations and deepfakes



Image Manipulation Detection and Localization



Video Manipulation Detection



Deepfakes (GAN) Detection GAN (Generative Adversarial Network)

Details at https://mfc.nist.gov



Ongoing Effort



Experiment design and data collection

- Synthetic (GAN-based) data generation
- Comparable real-world data collection



Web-based leaderboard

• Support simplicity & easy to participate



Interactive Dashboard

- Web-based data analysis (data contains rich metadata)
- Research direction for system improvement

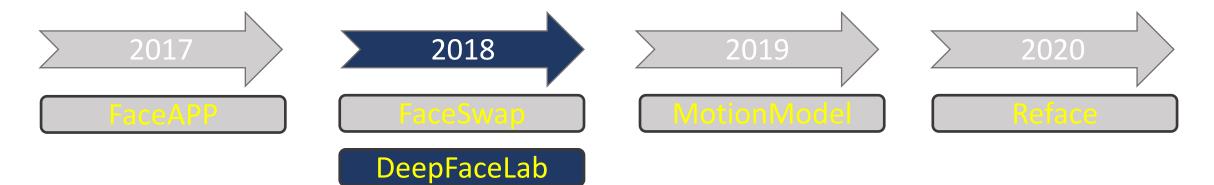
Deepfakes Generation

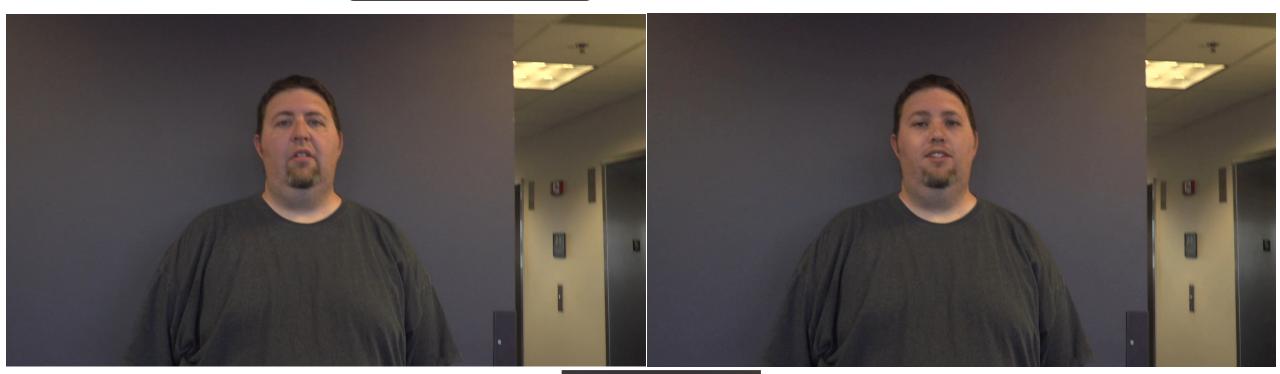
Completed Test

Continued Test

| Deepfakes Tools | Release | GAN Models | Release | GAN Datasets | Release |
|--------------------------|---------|-----------------------|---------|----------------------|---------|
| FaceApp | 2017 | PixelCNN, ProGAN | ~2017 | CityScapes, ADK20k | 2016 |
| Deepfakes FaceSwap | 2018 | SN-GAN, MMD-GAN, Glow | 2018 | CelebA-HQ, | 2017 |
| DeepFaceLab | 2018 | StyleGAN | 2019 | COCO-stuff, VGGFace2 | 2018 |
| First Order Motion Model | 2019 | FSGAN, StyleGAN2 | 2020 | FFHQ | 2019 |
| Reface | 2020 | StyleGAN3 | 2021 | AFHQ v2 | 2021 |

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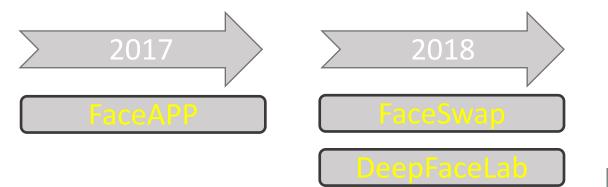


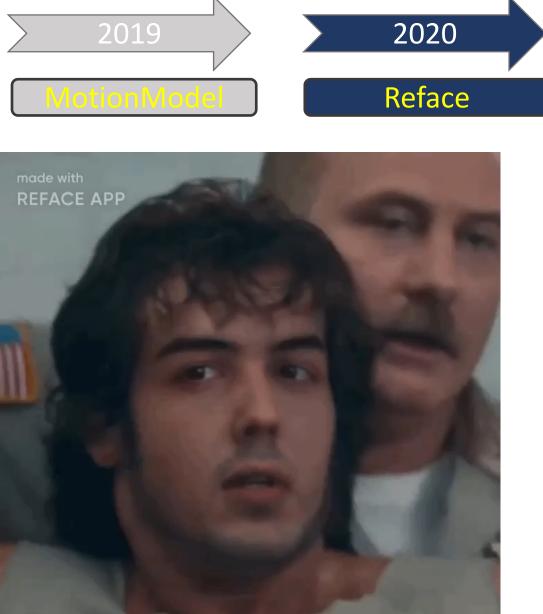


Base



Deepfaked

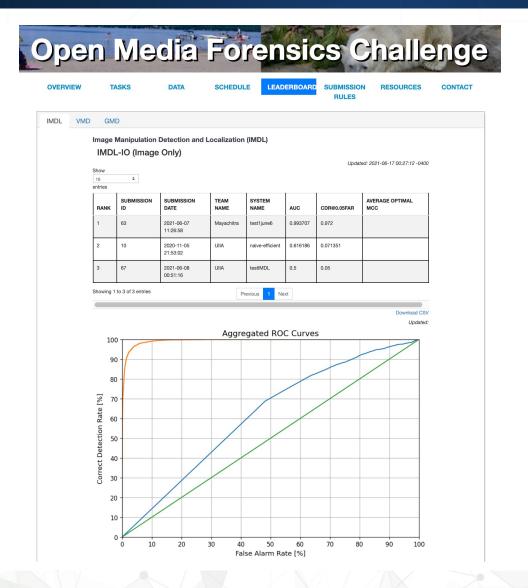






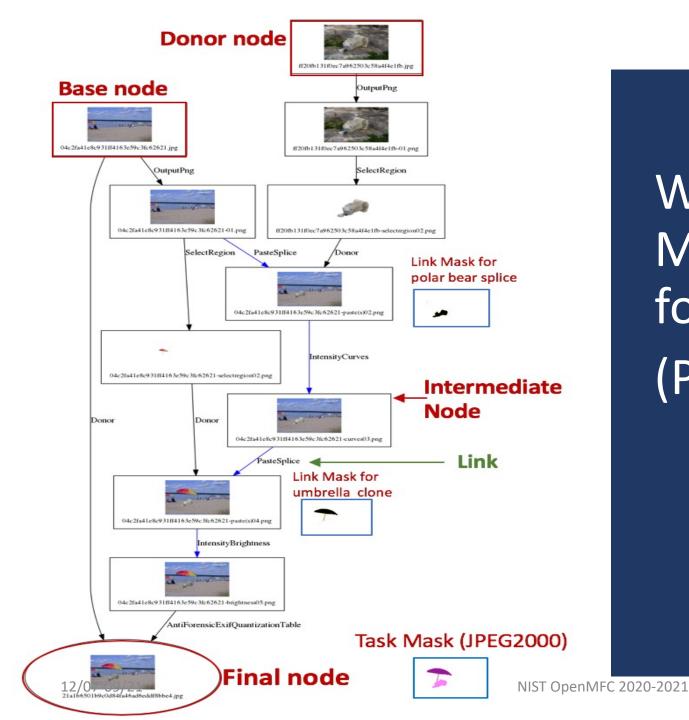
Deepfaked

Web-based Leaderboard



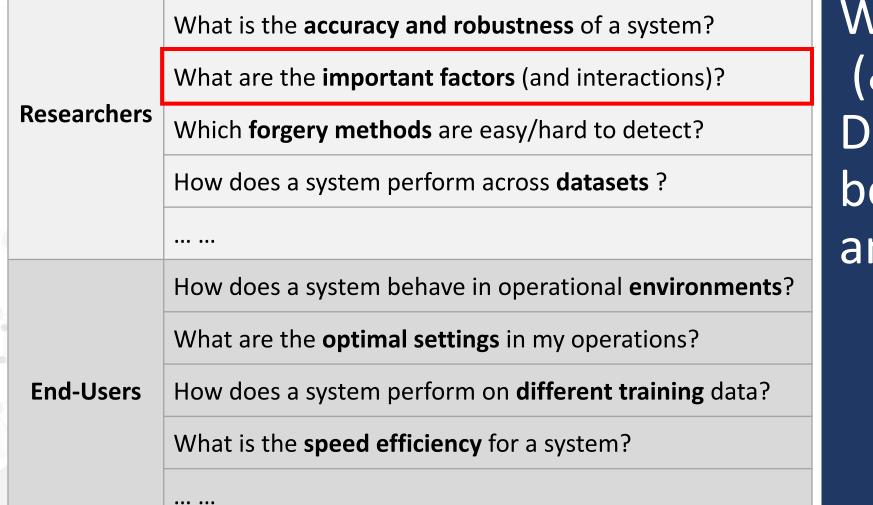
Quick Turnaround Leaderboard Evaluation

https://mfc.nist.gov



Web-based Media Level Analysis for Validation Set (Provenance)

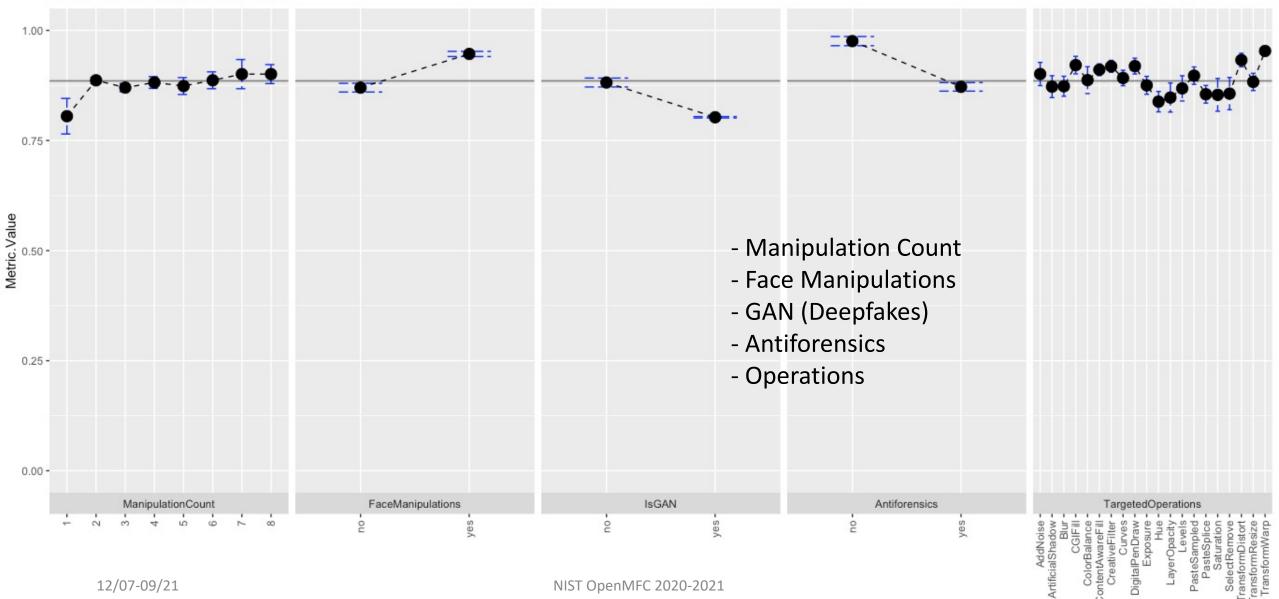
Interactive Dashboard



Web-based (& Interactive) Data Analysis for both researchers and end-users

Q: What are important factors that affect system performance?

Main Effects Plot with Error Bars





Our Vision



Expand to "Consequence Detection" beyond manipulation detection

- Systematically predicting motivation or intention behind the manipulations/deepfakes
- Categorization & Classification (e.g., violent incitement, vehicle accident)



Contribute to prevent disinformation and its threat



Build collaborations across sectors and engage community stakeholders

NIST AT A GLANCE

| | 3,400+ FEDERAL EMPLOYEES | 5 NOBEL PRIZES | Q 2 CAMPUSES GAITHERSBURG, MD [HQ] BOULDER, CO |
|--------------------------|---|---|--|
| | 3,500+ ASSOCIATES | 10 COLLABORATIVE INSTITUTES | 400+ BUSINESSES USING NIST FACILITIES |
| JAJA ManufacturingUSA | NATIONAL OFFICE COORDINATING 14 MANUFACTURING INSTITUTES | 51 MANUFACTURING EXTENSION PARTNERSHIP CENTERS | U.S. BALDRIGE PERFORMANCE EXCELLENCE PROGRAM |

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